

# Walmart Sales Forecasting using Machine Learning

April 18, 2023

## 0.1 Importing the necessary libraries

```
[1]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
%matplotlib inline
import warnings, re, joblib
warnings.filterwarnings('ignore')
from sklearn.model_selection import train_test_split, RandomizedSearchCV,
    RepeatedKFold, cross_val_score
from sklearn.linear_model import LinearRegression, PassiveAggressiveRegressor,
    SGDRegressor, ARDRegression, RidgeCV, LassoCV, ElasticNetCV,
    TweedieRegressor, HuberRegressor, RANSACRegressor
from sklearn.neighbors import KNeighborsRegressor
from sklearn.svm import SVR, LinearSVR, NuSVR
from sklearn.tree import DecisionTreeRegressor
from sklearn.ensemble import RandomForestRegressor, BaggingRegressor,
    ExtraTreesRegressor, GradientBoostingRegressor, AdaBoostRegressor,
    HistGradientBoostingRegressor, VotingRegressor
from sklearn.neural_network import MLPRegressor
from xgboost import XGBRegressor, XGBRFRegressor
from catboost import CatBoostRegressor
from lightgbm import LGBMRegressor
from scipy.stats import probplot
from sklearn.metrics import mean_absolute_percentage_error, mean_squared_error,
    r2_score
from sklearn.preprocessing import StandardScaler, FunctionTransformer,
    PowerTransformer, OrdinalEncoder
from yellowbrick.regressor import AlphaSelection, ResidualsPlot, PredictionError
from yellowbrick.features import Rank2D
import plotly.express as px
from sklearn.feature_selection import SelectKBest, SelectPercentile,
    SelectFromModel, f_regression, mutual_info_regression, RFECV,
    SequentialFeatureSelector
```

## 0.2 Loading the dataset

```
[2]: df = pd.read_csv('walmart-sales-dataset-of-45stores.csv')
df.head()
```

```
[2]:    Store      Date  Weekly_Sales  Holiday_Flag  Temperature  Fuel_Price \
0       1  05-02-2010     1643690.90          0        42.31     2.572
1       1  12-02-2010     1641957.44          1        38.51     2.548
2       1  19-02-2010     1611968.17          0        39.93     2.514
3       1  26-02-2010     1409727.59          0        46.63     2.561
4       1  05-03-2010     1554806.68          0        46.50     2.625

           CPI  Unemployment
0  211.096358        8.106
1  211.242170        8.106
2  211.289143        8.106
3  211.319643        8.106
4  211.350143        8.106
```

## 0.3 Understanding and assessing the shape and structure of the dataset

```
[3]: df.shape
```

```
[3]: (6435, 8)
```

```
[4]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 6435 entries, 0 to 6434
Data columns (total 8 columns):
 #   Column            Non-Null Count  Dtype  
--- 
 0   Store              6435 non-null   int64  
 1   Date               6435 non-null   object  
 2   Weekly_Sales       6435 non-null   float64
 3   Holiday_Flag       6435 non-null   int64  
 4   Temperature        6435 non-null   float64
 5   Fuel_Price         6435 non-null   float64
 6   CPI                6435 non-null   float64
 7   Unemployment       6435 non-null   float64
dtypes: float64(5), int64(2), object(1)
memory usage: 402.3+ KB
```

```
[5]: df.describe()
```

```
[5]:      Store  Weekly_Sales  Holiday_Flag  Temperature  Fuel_Price \
count  6435.000000  6.435000e+03  6435.000000  6435.000000  6435.000000
mean   23.000000  1.046965e+06      0.069930    60.663782    3.358607
```

```
      std    12.988182  5.643666e+05     0.255049   18.444933   0.459020  
      min     1.000000  2.099862e+05     0.000000  -2.060000   2.472000  
      25%    12.000000  5.533501e+05     0.000000   47.460000   2.933000  
      50%    23.000000  9.607460e+05     0.000000   62.670000   3.445000  
      75%    34.000000  1.420159e+06     0.000000   74.940000   3.735000  
      max    45.000000  3.818686e+06     1.000000  100.140000  4.468000
```

	CPI	Unemployment
count	6435.000000	6435.000000
mean	171.578394	7.999151
std	39.356712	1.875885
min	126.064000	3.879000
25%	131.735000	6.891000
50%	182.616521	7.874000
75%	212.743293	8.622000
max	227.232807	14.313000

```
[6]: df.isna().sum()
```

```
[6]: Store          0  
Date           0  
Weekly_Sales   0  
Holiday_Flag   0  
Temperature    0  
Fuel_Price     0  
CPI            0  
Unemployment   0  
dtype: int64
```

```
[7]: df.duplicated().sum()
```

```
[7]: 0
```

## 0.4 Exploratory Data Analysis (EDA)

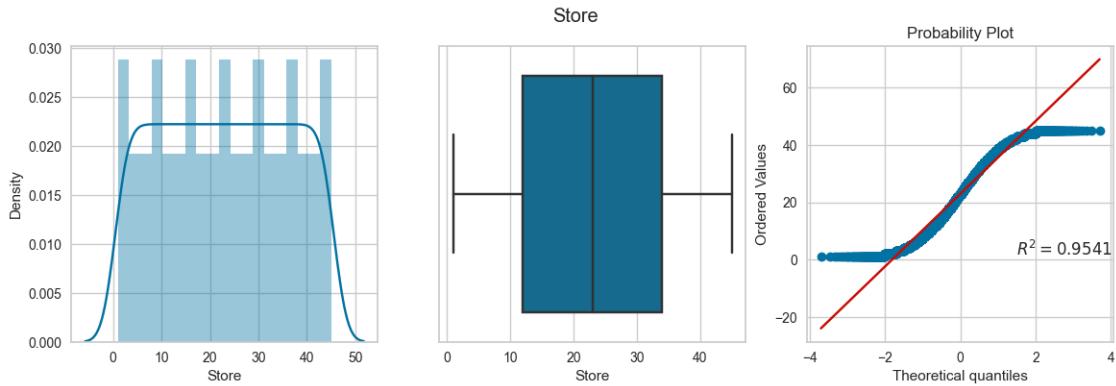
### 0.4.1 Univariate Analysis

```
[8]: for col in df.select_dtypes(np.number).columns:  
    plt.figure(figsize=(14,4))  
    print(f"Skewness of {col}:",df[col].skew())  
    print(f"Kurtosis of {col}:",df[col].kurtosis())  
    plt.subplot(131)  
    sns.distplot(df[col])  
    plt.subplot(132)  
    sns.boxplot(df[col])  
    plt.subplot(133)  
    probplot(df[col],rvalue=True,plot=plt,dist='norm')  
    plt.suptitle(col)
```

```
plt.show();
```

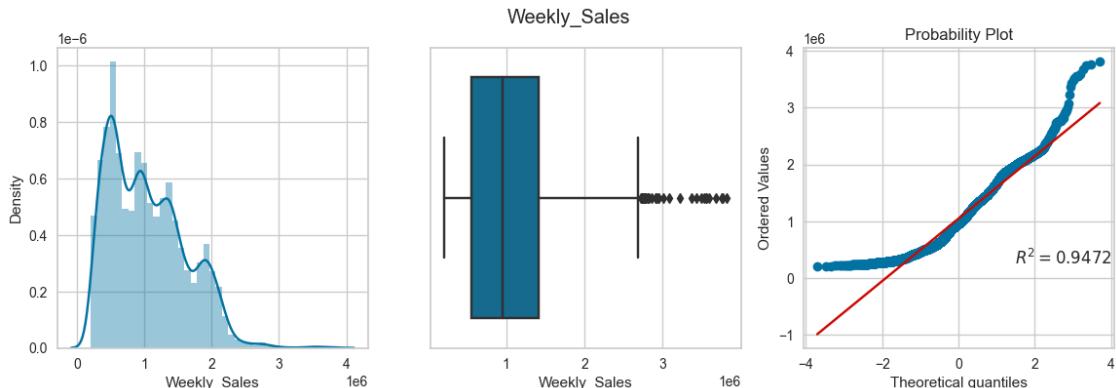
Skewness of Store: 0.0

Kurtosis of Store: -1.2011866346083275



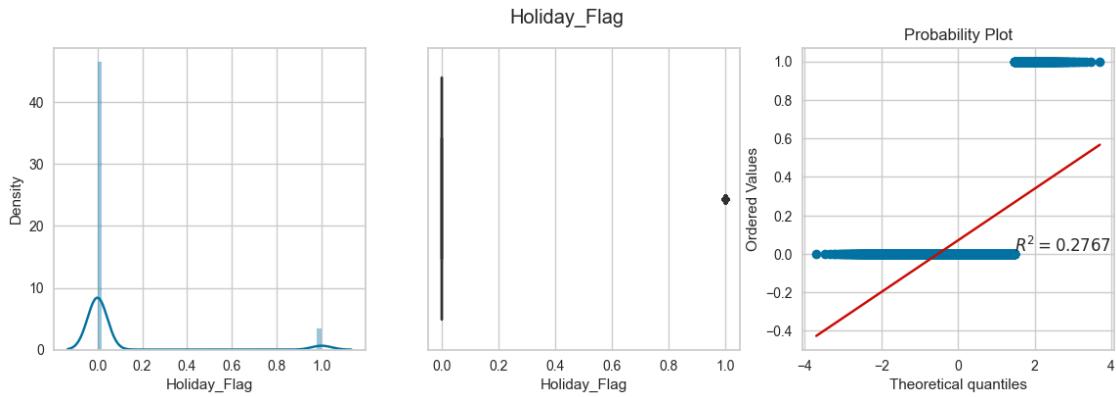
Skewness of Weekly\_Sales: 0.6683617974864524

Kurtosis of Weekly\_Sales: 0.05314092741722032

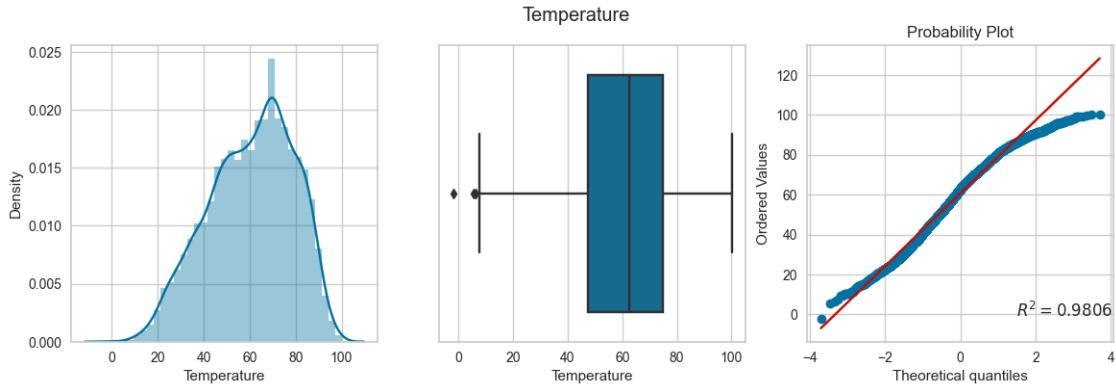


Skewness of Holiday\_Flag: 3.3734986714578485

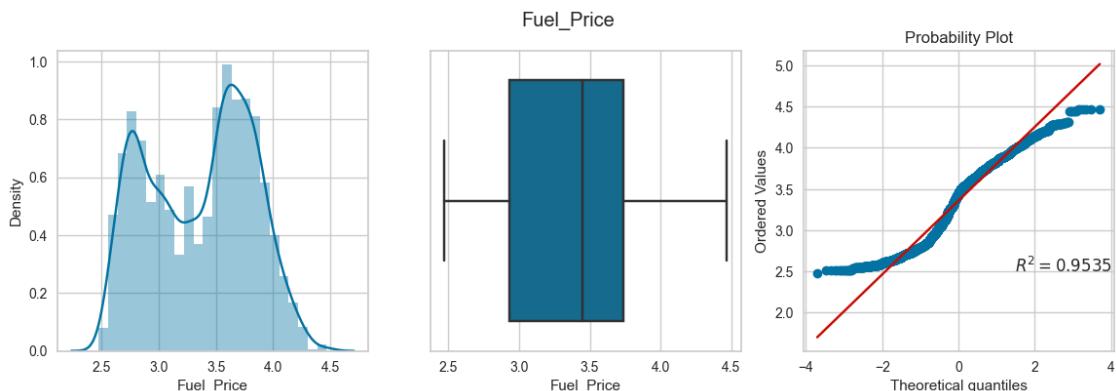
Kurtosis of Holiday\_Flag: 9.383409556463409



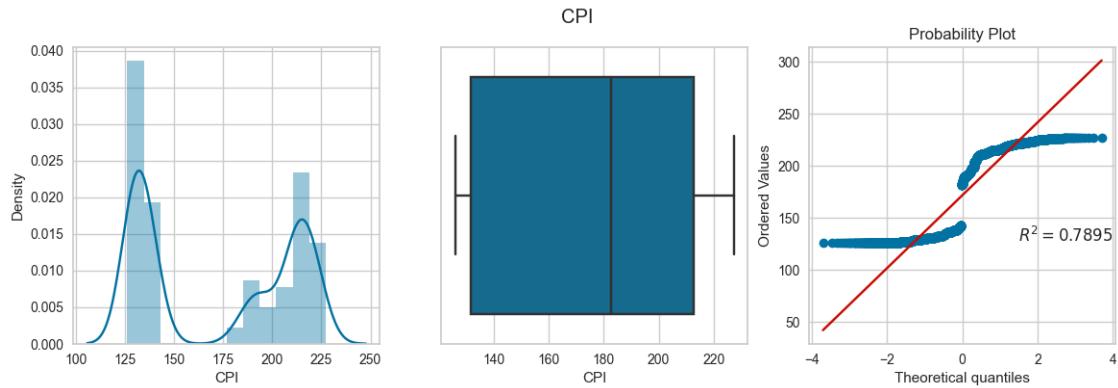
Skewness of Temperature: -0.3367676011075799  
 Kurtosis of Temperature: -0.6128009588453383



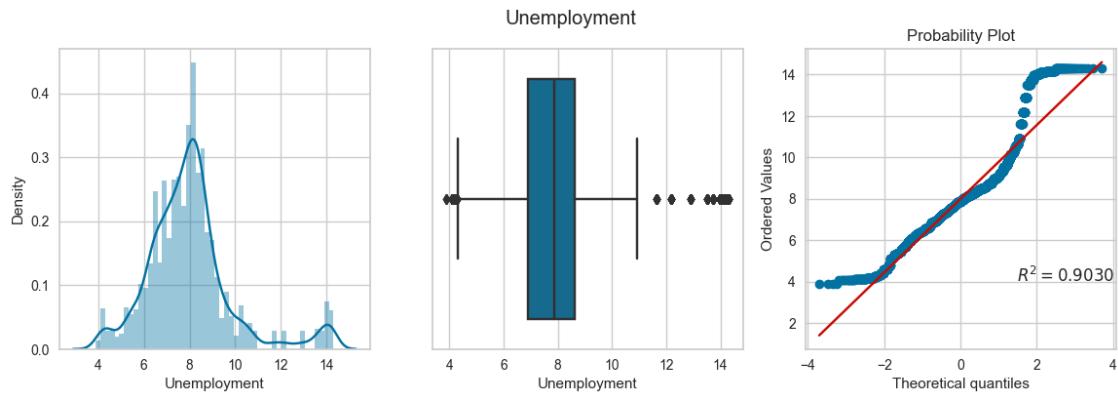
Skewness of Fuel\_Price: -0.09615830011865549  
 Kurtosis of Fuel\_Price: -1.1773777964906604



Skewness of CPI: 0.06349184988549494  
 Kurtosis of CPI: -1.8398133421838165

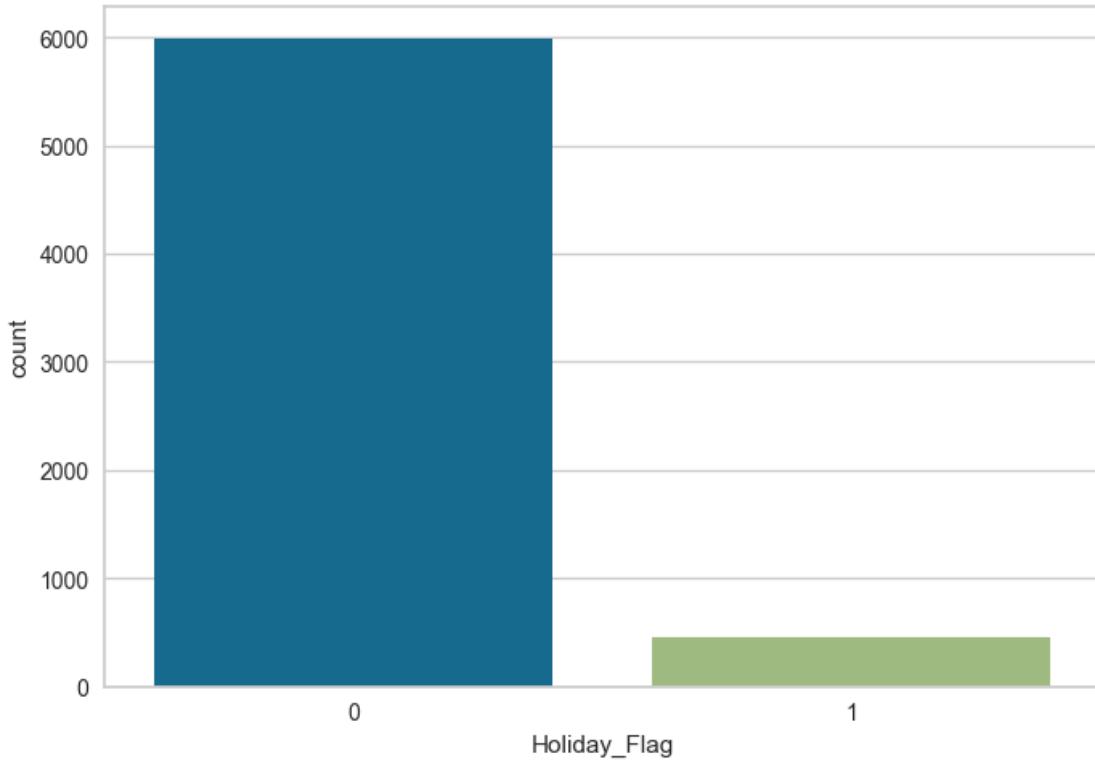


Skewness of Unemployment: 1.1881439334843265  
 Kurtosis of Unemployment: 2.639711784234234



Normal Distribution: Store Left-Skewed Distribution: Temperature Right-Skewed Distribution: Weekly\_Sales, Unemployment, Fuel\_Price Bimodal Distribution: CPI Outlier Columns: Weekly\_Sales, Temperature, Unemployment

[9]: `sns.countplot(df.Holiday_Flag);`



There are very few holidays across the given time period.

#### 0.4.2 Bivariate Analysis

```
[10]: df.corr()['Weekly_Sales'].sort_values(ascending=False)[1:]
```

```
[10]: Holiday_Flag      0.036891
Fuel_Price        0.009464
Temperature     -0.063810
CPI            -0.072634
Unemployment   -0.106176
Store          -0.335332
Name: Weekly_Sales, dtype: float64
```

```
[11]: px.scatter(df,x='CPI',y='Weekly_Sales',color='Holiday_Flag',title='2D Scatter ↴Plot of CPI and Weekly Sales')
```

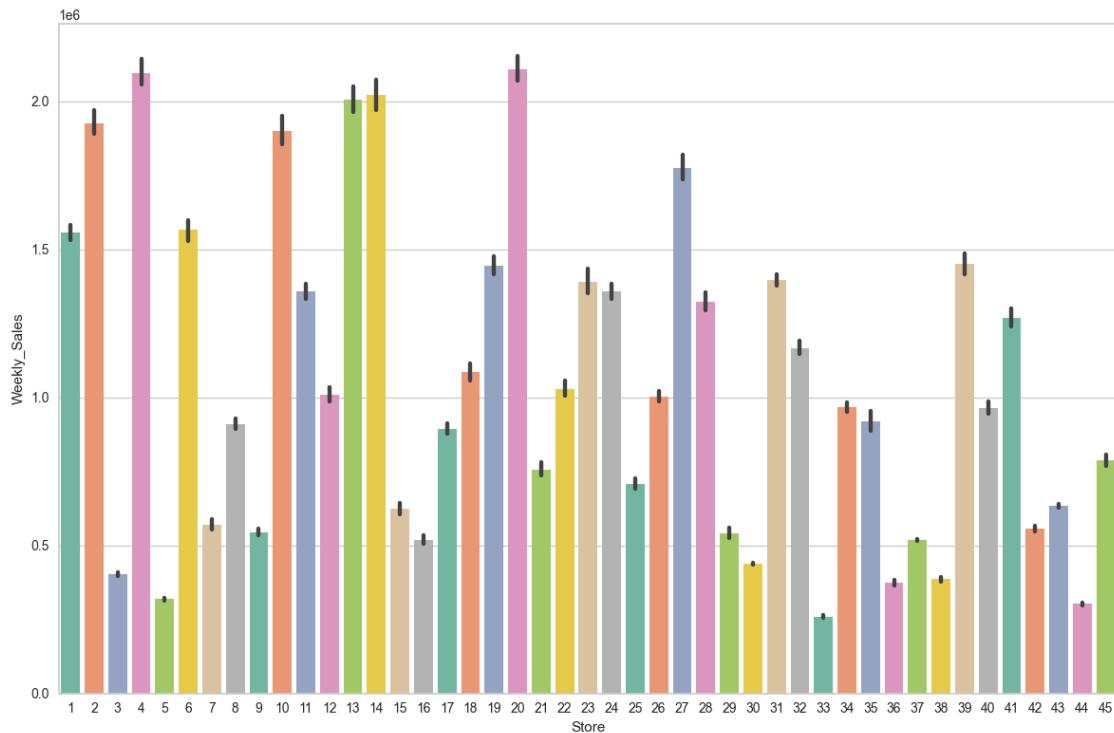
```
[12]: px.scatter(df,x='Unemployment',y='Weekly_Sales',color='Holiday_Flag',title='2D Scatter Plot of Unemployment and Weekly Sales')
```

There is a very weak negative correlation between unemployment and weekly sales.

```
[13]: px.scatter(df,x='Temperature',y='Weekly_Sales',color='Holiday_Flag',title='2D  
↳Scatter Plot of Temperature and Weekly Sales')
```

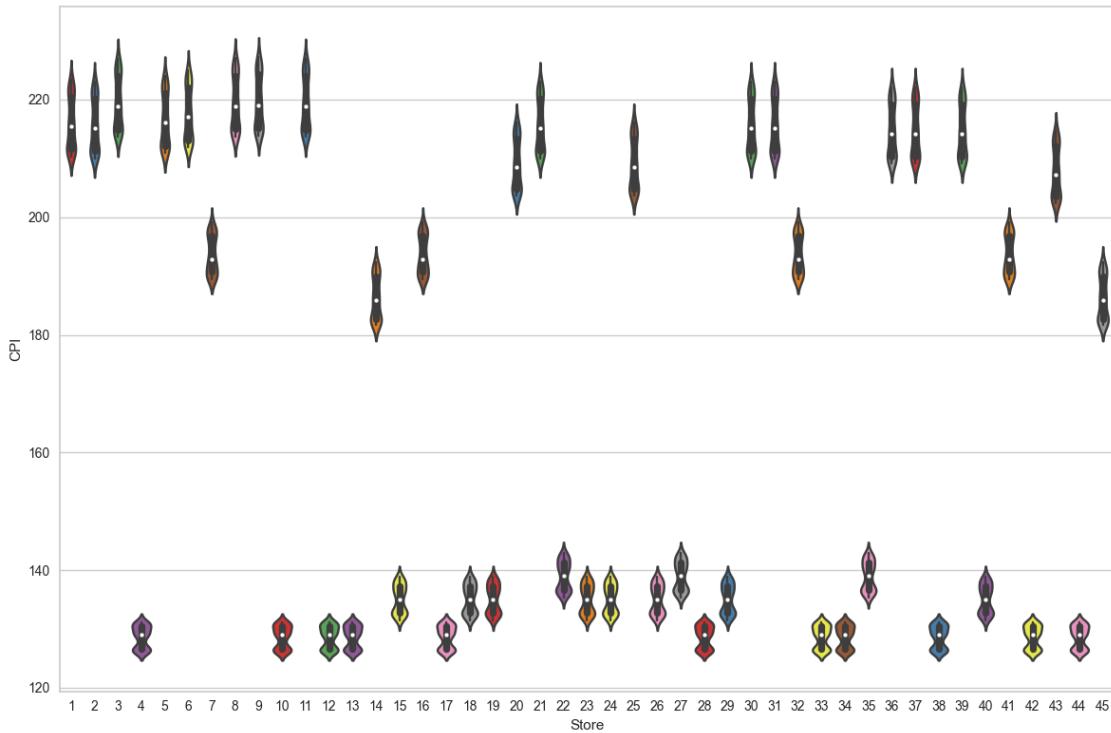
There are burgeoning weekly sales during the sunny days and clear weather when temperature ranges between 25 deg C to 45 deg C.

```
[14]: plt.figure(figsize=(12,8))
sns.barplot(x='Store',y='Weekly_Sales',data=df,palette='Set2')
plt.tight_layout();
```



Stores 4 and 20 have had the most flourishing sales whereas store 33 has had really poor sales.

```
[15]: plt.figure(figsize=(12,8))
sns.violinplot(x='Store',y='CPI',data=df,palette='Set1')
plt.tight_layout();
```



Stores 3,8,9 and 11 have the highest Consumer Price Index (CPI) i.e. they have raised the selling price by a significant amount in the past few years. In simple words, these stores have undergone the most inflation from 2010 to 2012. On the other hand, there are many stores which have the lowest CPI such as stores 4,10,12,13,17,28,33,34,38,42 and 44.

```
[16]: holidays = df[df.Holiday_Flag == 1]
workdays = df[df.Holiday_Flag == 0]
```

```
[17]: workdays.describe()
```

	Store	Weekly_Sales	Holiday_Flag	Temperature	Fuel_Price
count	5985.000000	5.985000e+03	5985.0	5985.000000	5985.000000
mean	23.000000	1.041256e+06	0.0	61.448124	3.368467
std	12.988258	5.589574e+05	0.0	18.076892	0.461158
min	1.000000	2.099862e+05	0.0	-2.060000	2.472000
25%	12.000000	5.513784e+05	0.0	48.730000	2.932000
50%	23.000000	9.562112e+05	0.0	63.560000	3.475000
75%	34.000000	1.414344e+06	0.0	75.320000	3.744000
max	45.000000	3.818686e+06	0.0	100.140000	4.468000

	CPI	Unemployment
count	5985.000000	5985.000000
mean	171.601725	7.993514
std	39.364794	1.875664

```

min      126.064000      3.879000
25%     131.784000      6.891000
50%     182.622509      7.874000
75%     212.861131      8.595000
max     227.232807     14.313000

```

[18]: `holidays.describe()`

```

[18]:          Store  Weekly_Sales  Holiday_Flag  Temperature  Fuel_Price  \
count   450.000000  4.500000e+02      450.0    450.000000  450.000000
mean    23.000000  1.122888e+06      1.0     50.232044   3.227464
std     13.001627  6.276849e+05      0.0     20.071118   0.407934
min     1.000000  2.153592e+05      1.0     10.240000   2.513000
25%    12.000000  5.758655e+05      1.0     33.367500   2.943000
50%    23.000000  1.018538e+06      1.0     47.870000   3.179000
75%    34.000000  1.555213e+06      1.0     67.335000   3.546000
max    45.000000  3.004702e+06      1.0     96.220000   4.124000

                  CPI  Unemployment
count   450.000000      450.000000
mean    171.268092      8.074127
std     39.291449      1.879293
min     126.114581      4.077000
25%    131.586613      7.015000
50%    182.598178      7.890000
75%    211.760416      8.625000
max    226.210354     14.313000

```

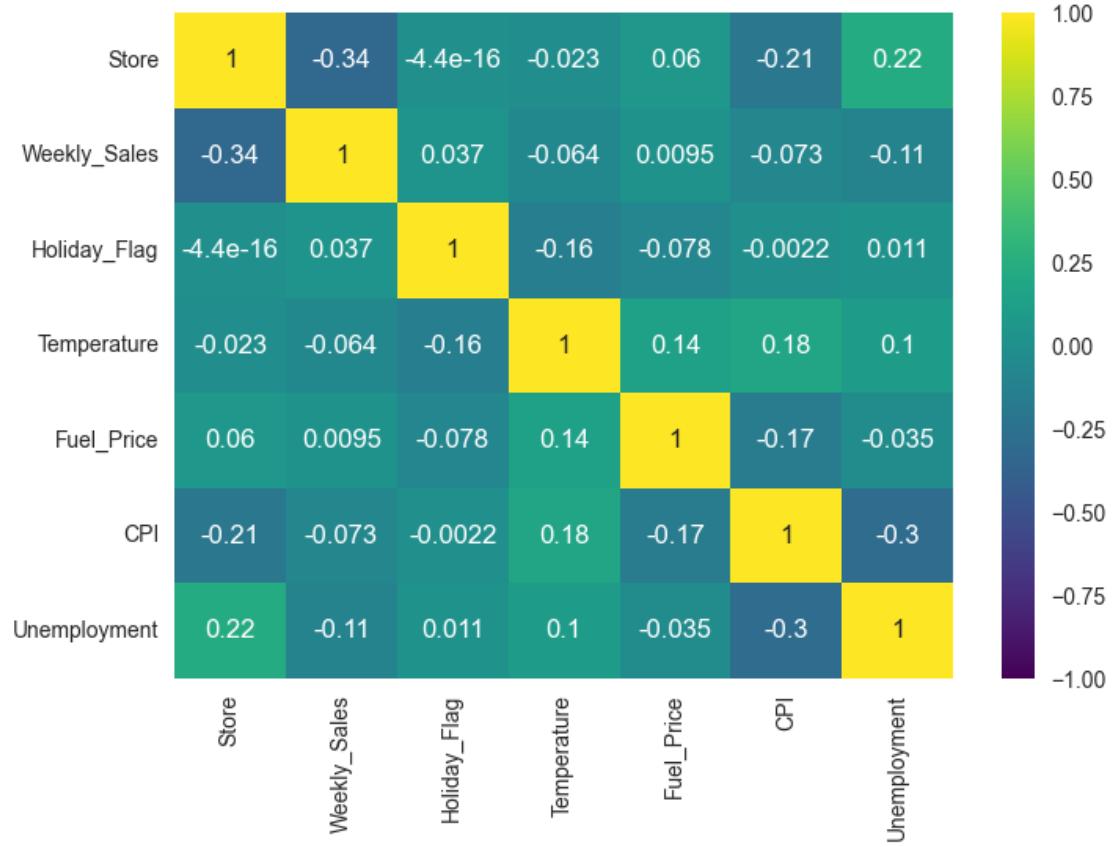
Holidays have slightly more weekly sales as compared to normal days. This may be due to the festive seasons when a large number of consumers visit the retail stores for purchasing their desired goods, products and services at reasonable discounted rates. Temperatures are relatively lower in the holiday season in comparison to the normal working season.

#### 0.4.3 Multivariate Analysis

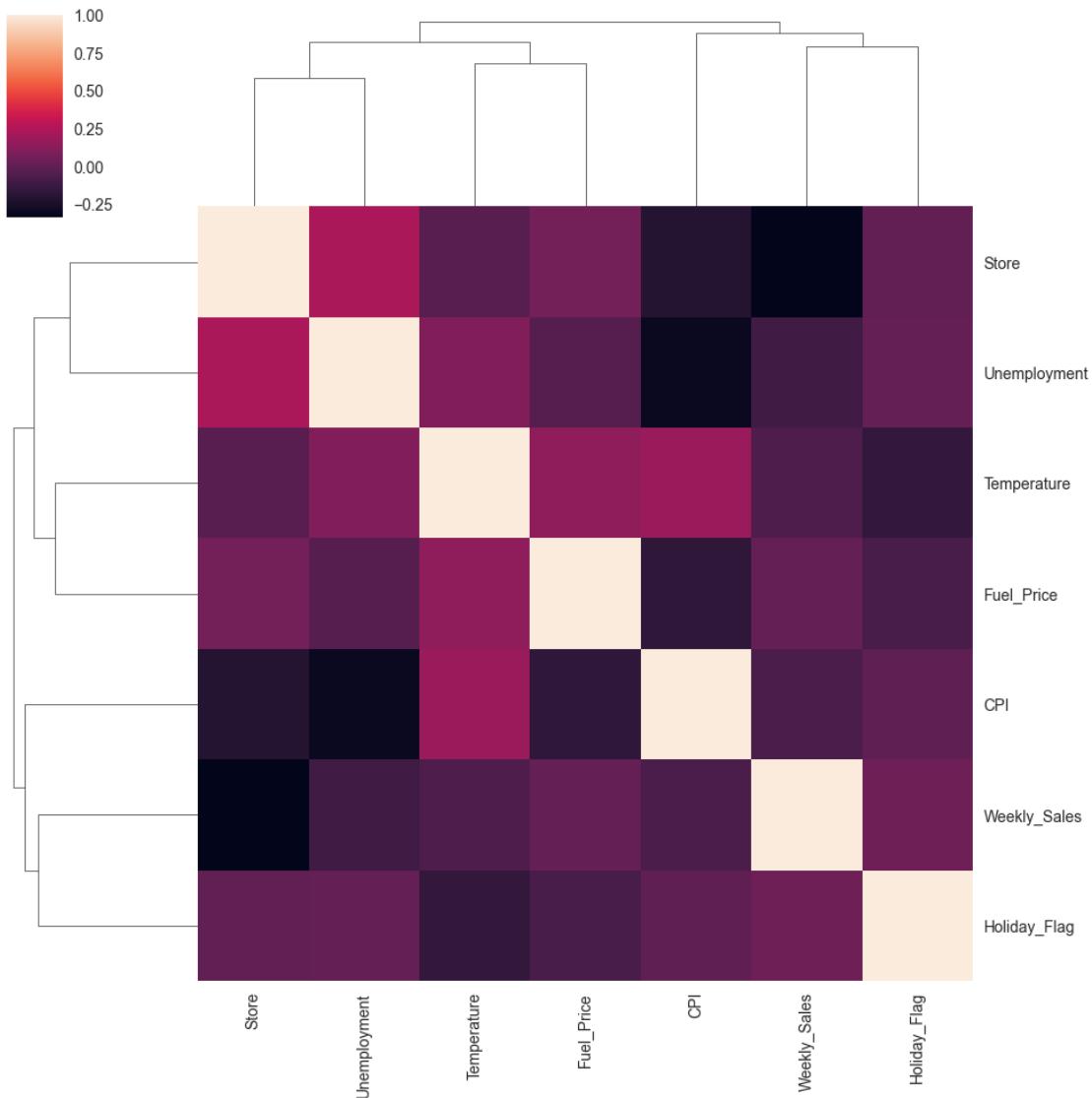
[19]: `fig = px.  
 scatter_3d(df,x='CPI',y='Weekly_Sales',z='Fuel_Price',color='Holiday_Flag',title='3D  
 Scatter Plot of CPI and Weekly Sales')  
 fig.show()`

Fuel price is generally lower during the holidays.

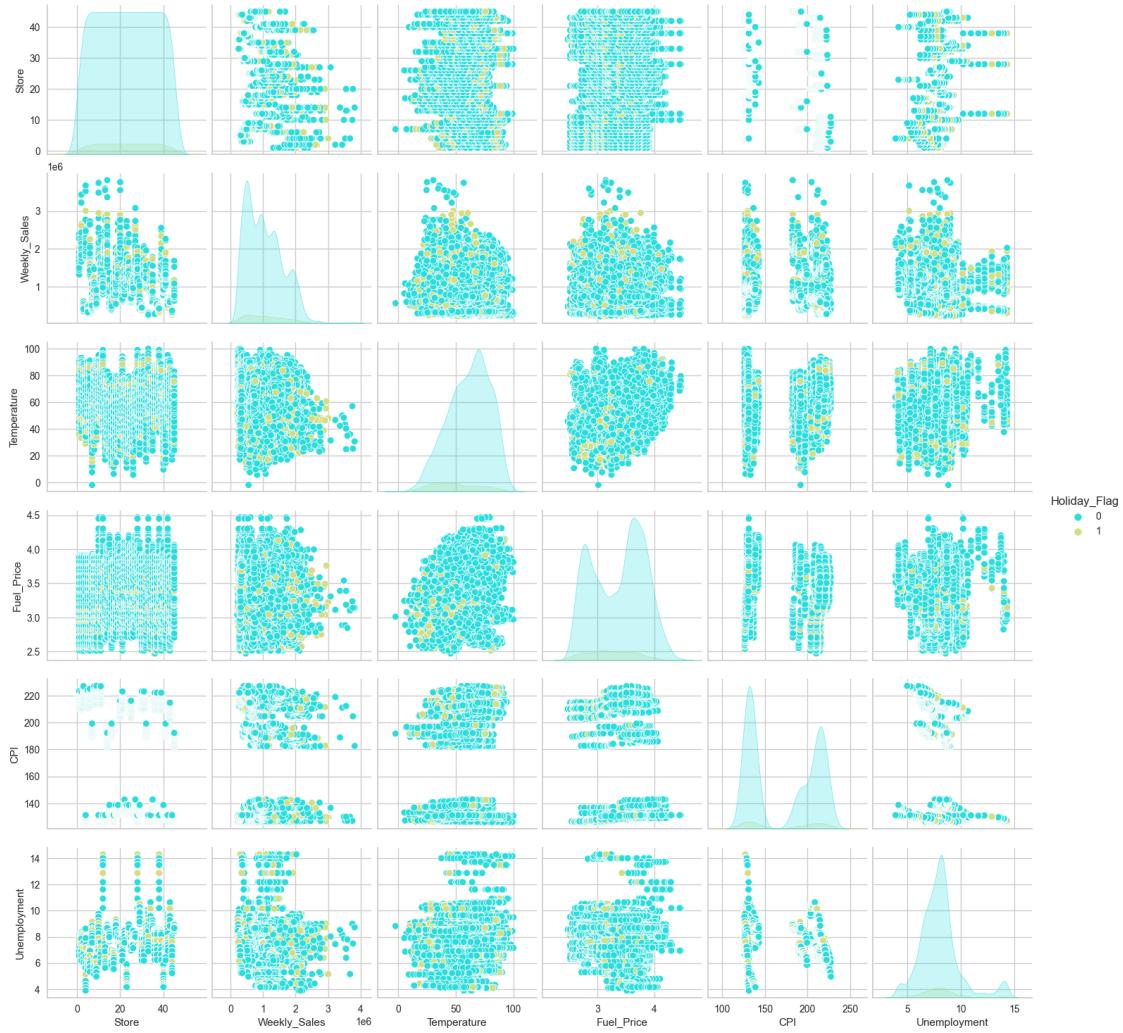
[20]: `sns.heatmap(df.corr(),annot=True,cmap='viridis',vmin=-1.0,vmax=1.0);`



```
[21]: sns.clustermap(df.corr());
```



```
[22]: sns.pairplot(df,hue='Holiday_Flag',palette='rainbow');
```



## 0.5 Feature Engineering

### 0.5.1 Outlier Treatment

```
[23]: def remove_outliers(col):
    lower_limit, upper_limit = df[col].quantile([0.25,0.75])
    IQR = upper_limit - lower_limit
    lower_whisker = lower_limit - 1.5 * IQR
    upper_whisker = upper_limit + 1.5 * IQR
    return np.where(df[col]>upper_whisker,upper_whisker,np.
    ↪where(df[col]<lower_whisker,lower_whisker,df[col]))
```

```
[24]: outlier_cols = ['Weekly_Sales', 'Temperature', 'Unemployment']

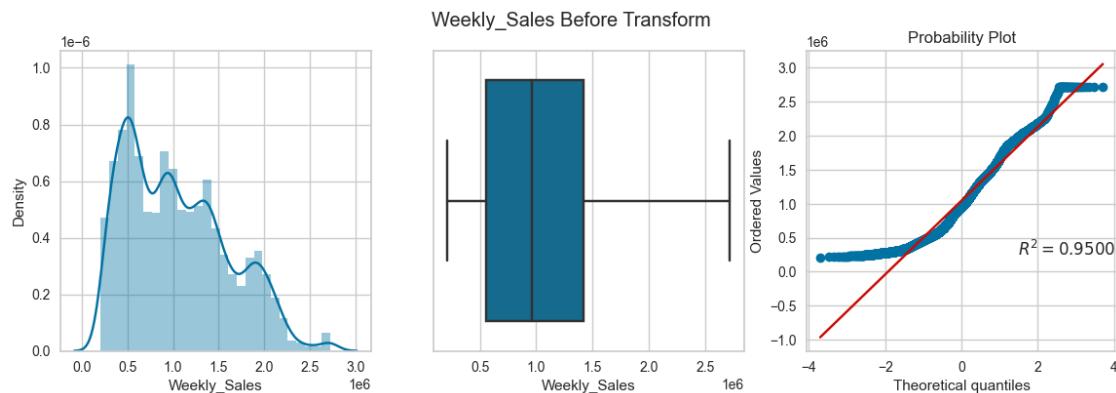
for col in outlier_cols:
    df[col] = remove_outliers(col)
```

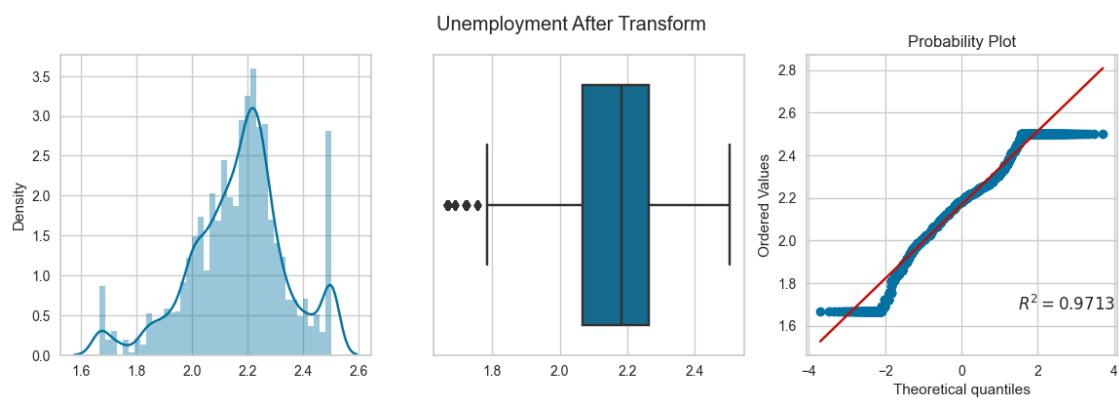
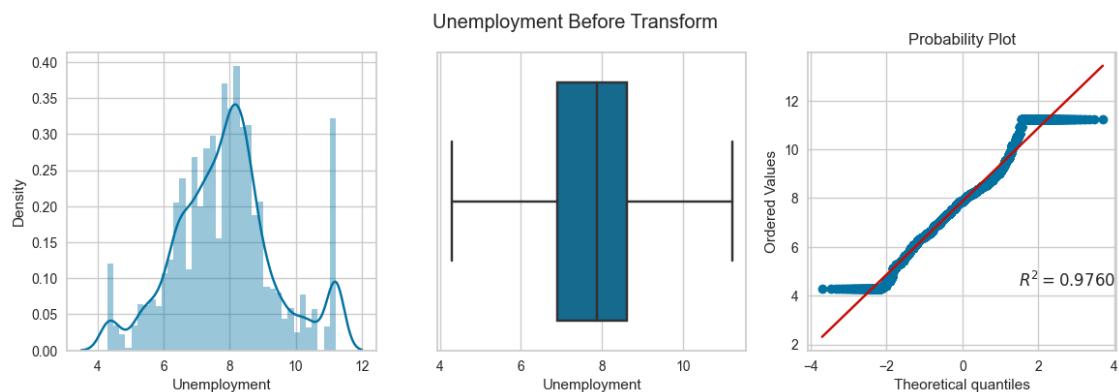
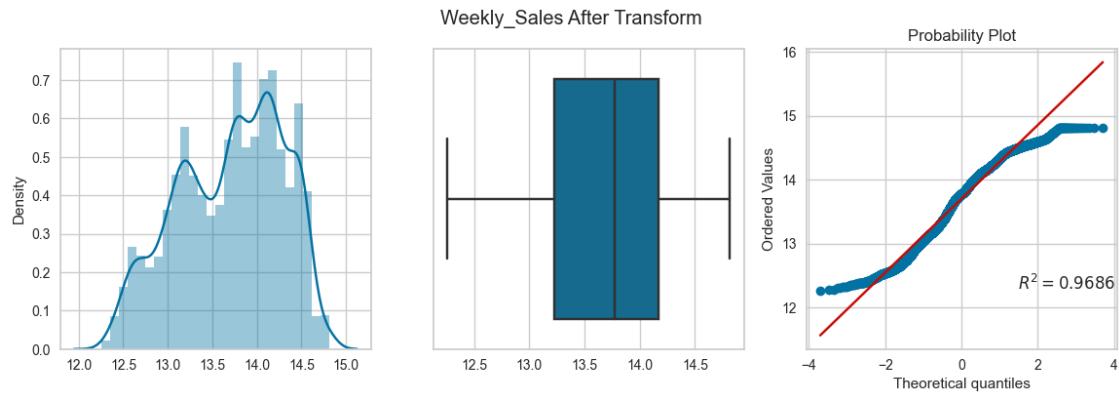
## 0.5.2 Feature Transformation

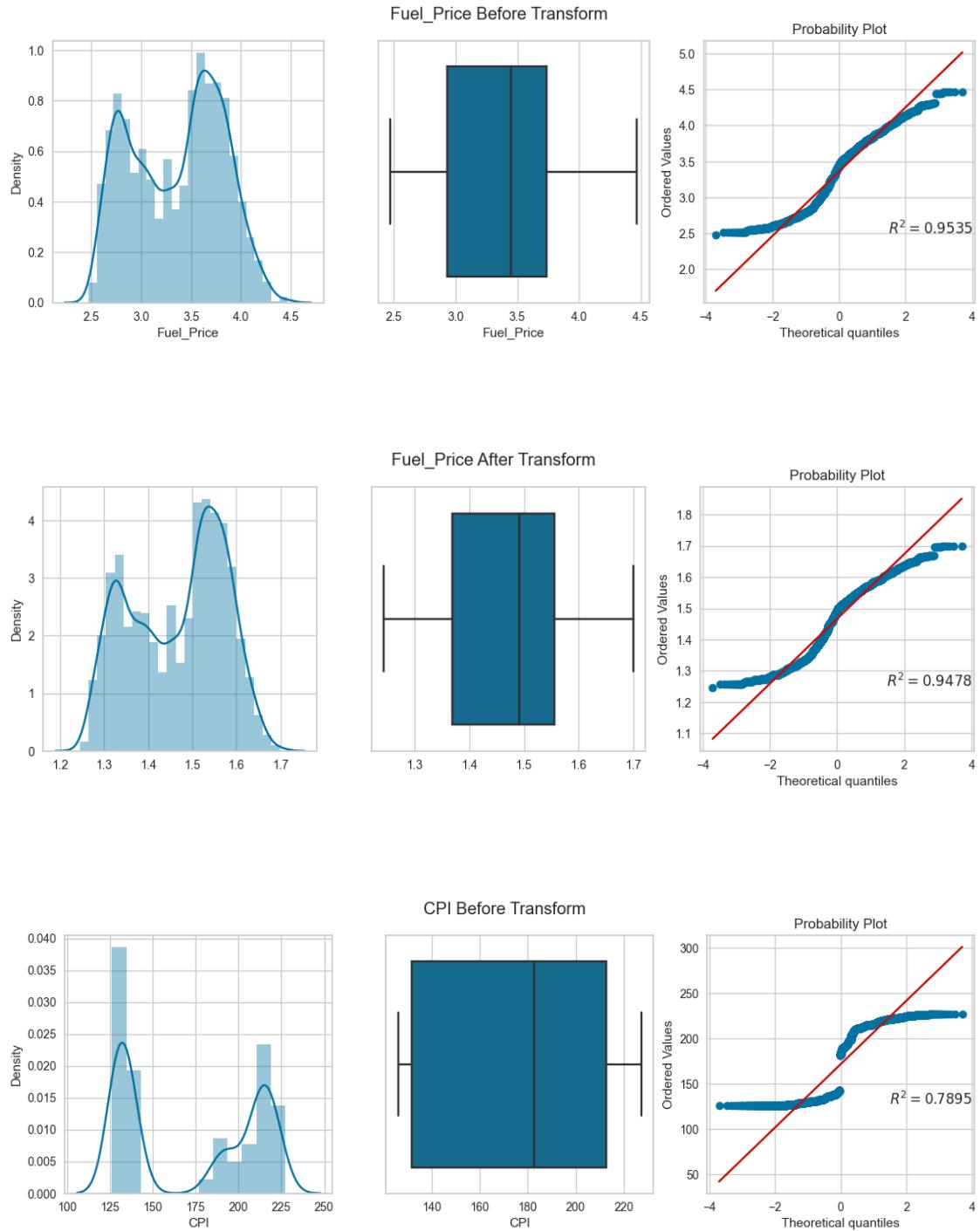
```
[25]: def apply_transform(transform,col):
    plt.figure(figsize=(14,4))
    plt.subplot(131)
    sns.distplot(df[col])
    plt.subplot(132)
    sns.boxplot(df[col])
    plt.subplot(133)
    probplot(df[col],rvalue=True,dist='norm',plot=plt)
    plt.suptitle(f'{col} Before Transform')
    plt.show()
    col_tf = transform.fit_transform(df[[col]])
    col_tf = np.array(col_tf).reshape(col_tf.shape[0])
    plt.figure(figsize=(14,4))
    plt.subplot(131)
    sns.distplot(col_tf)
    plt.subplot(132)
    sns.boxplot(col_tf)
    plt.subplot(133)
    probplot(col_tf,rvalue=True,dist='norm',plot=plt)
    plt.suptitle(f'{col} After Transform')
    plt.show();
```

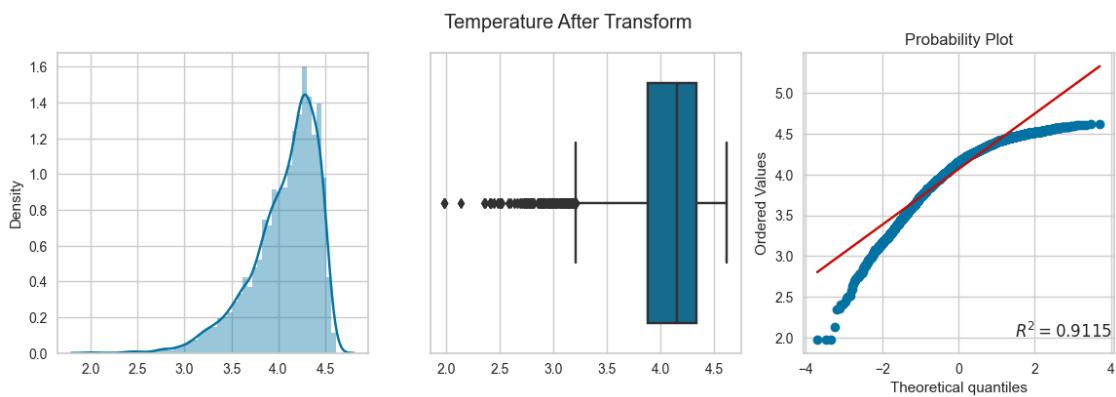
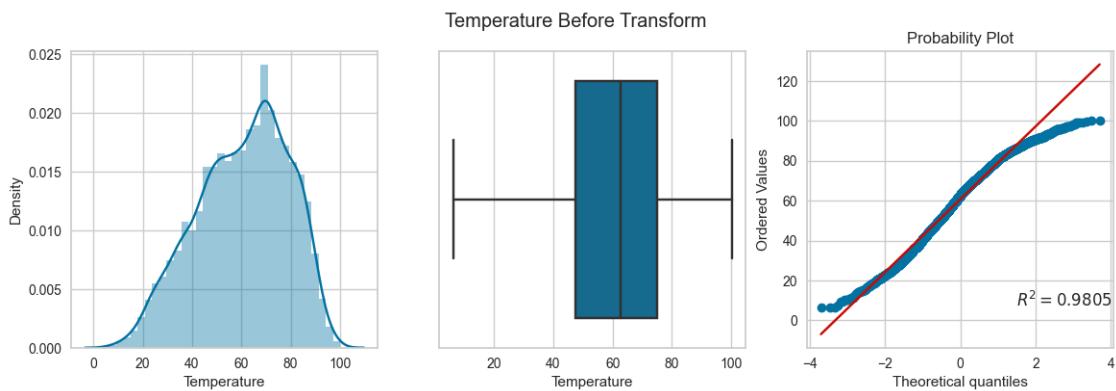
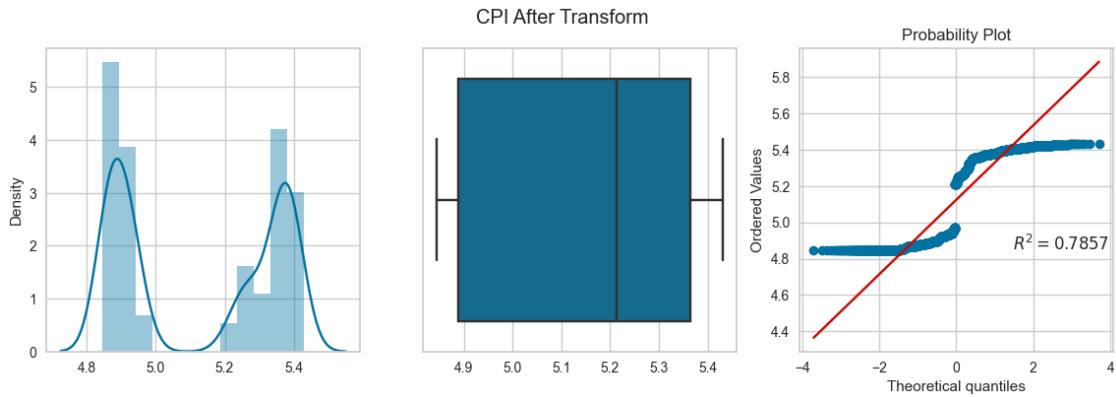
```
[26]: skewed_cols = ['Weekly_Sales', 'Unemployment', 'Fuel_Price', 'CPI', 'Temperature']

for col in skewed_cols:
    apply_transform(FunctionTransformer(np.log1p),col)
```

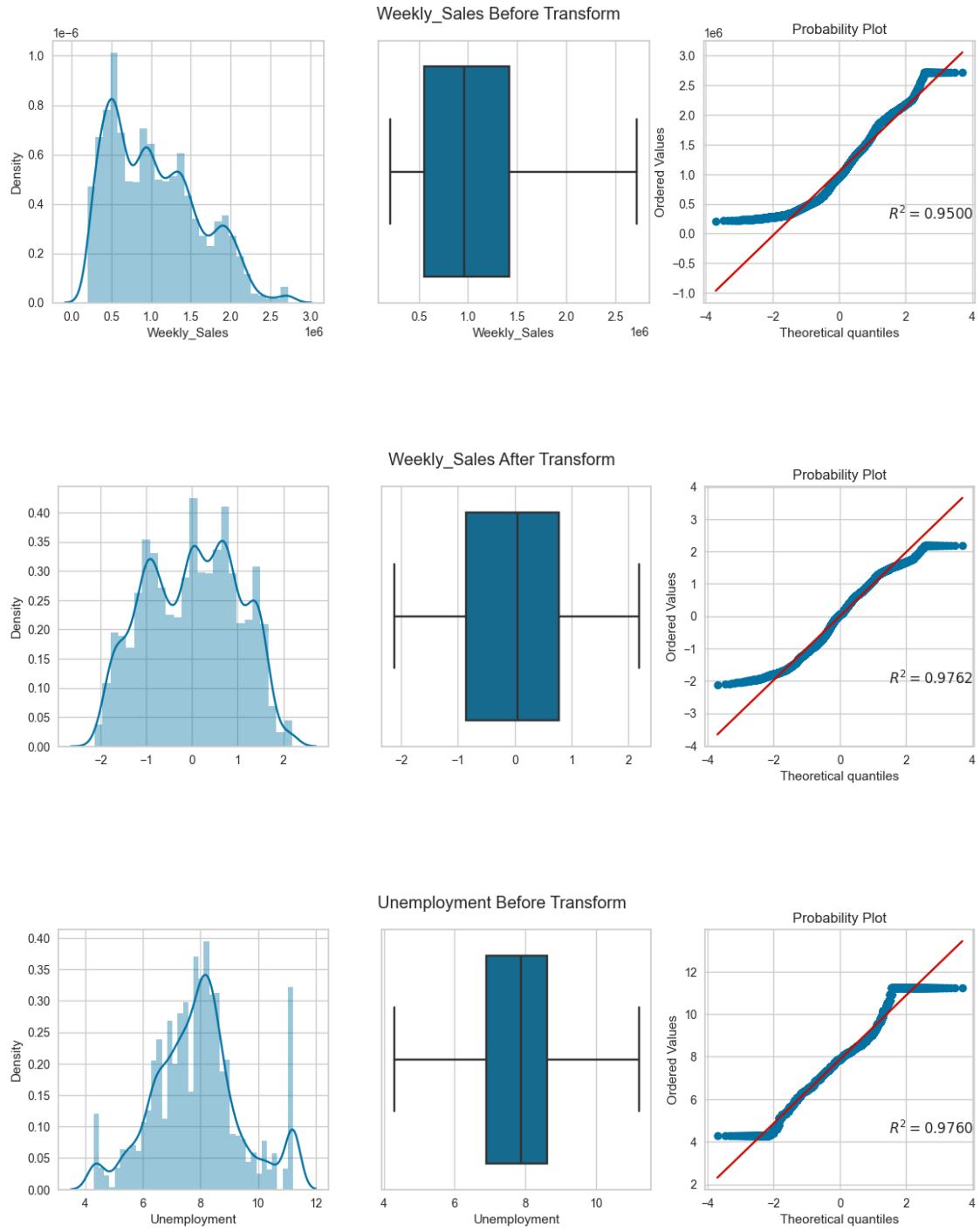




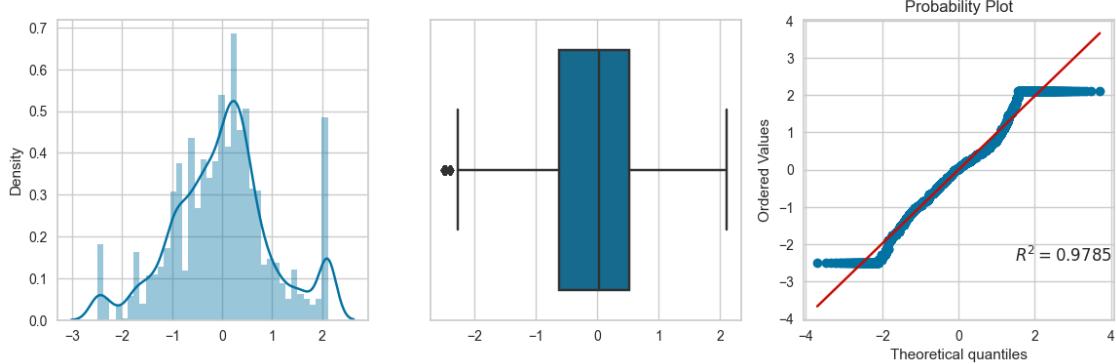




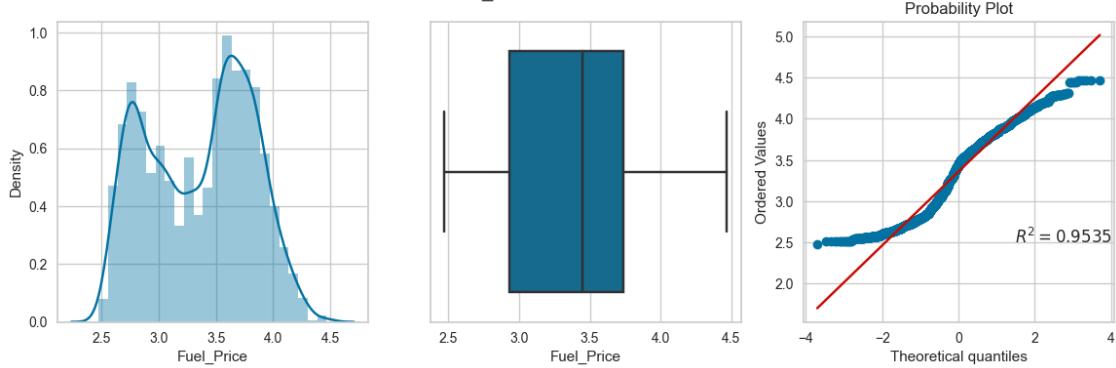
```
[27]: for col in skewed_cols:
    apply_transform(PowerTransformer(), col)
```



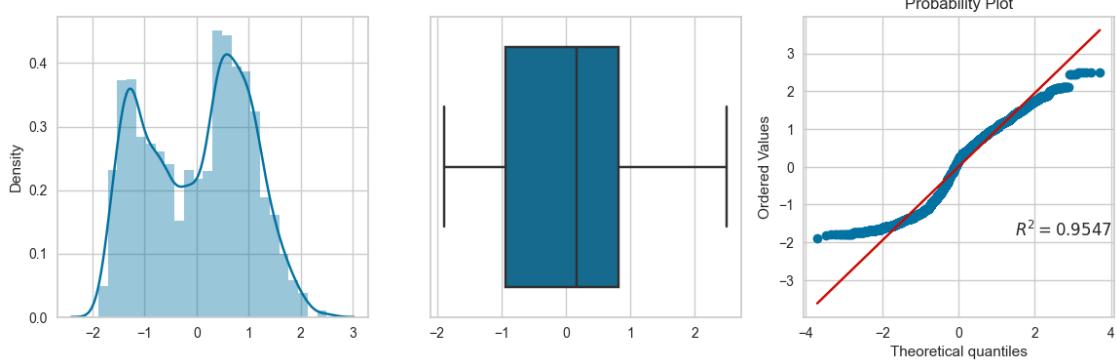
Unemployment After Transform

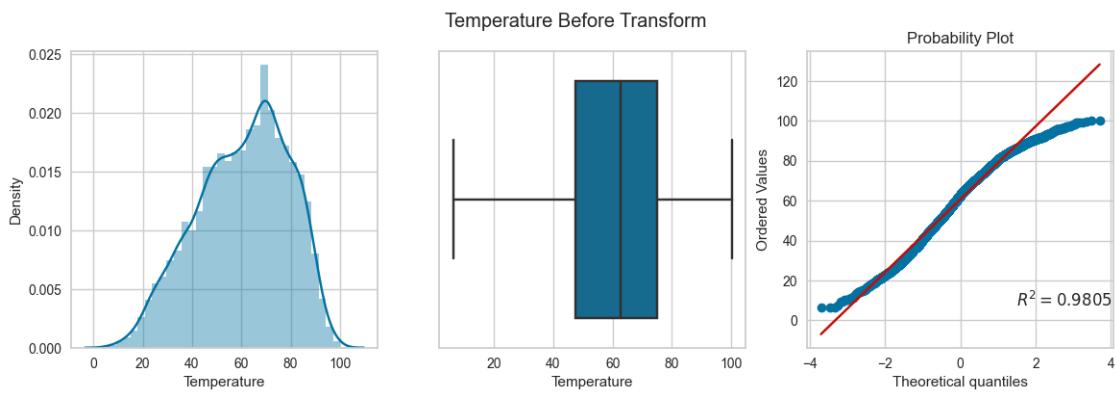
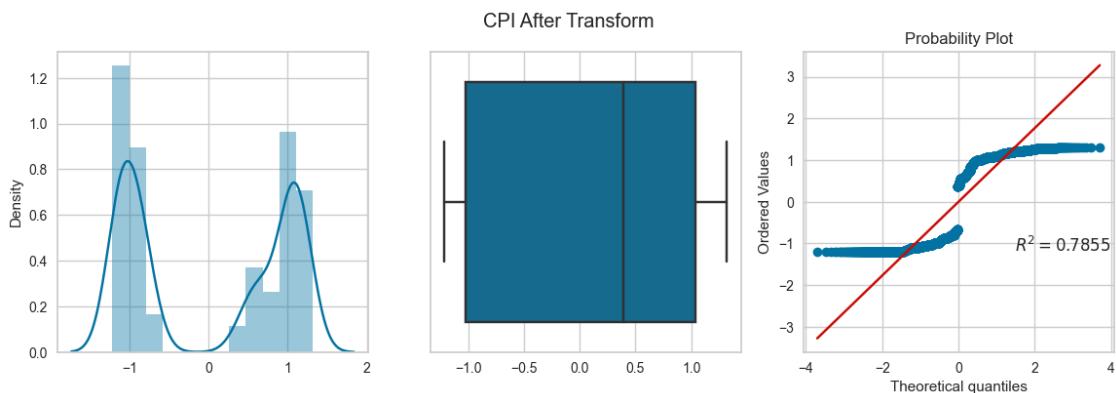
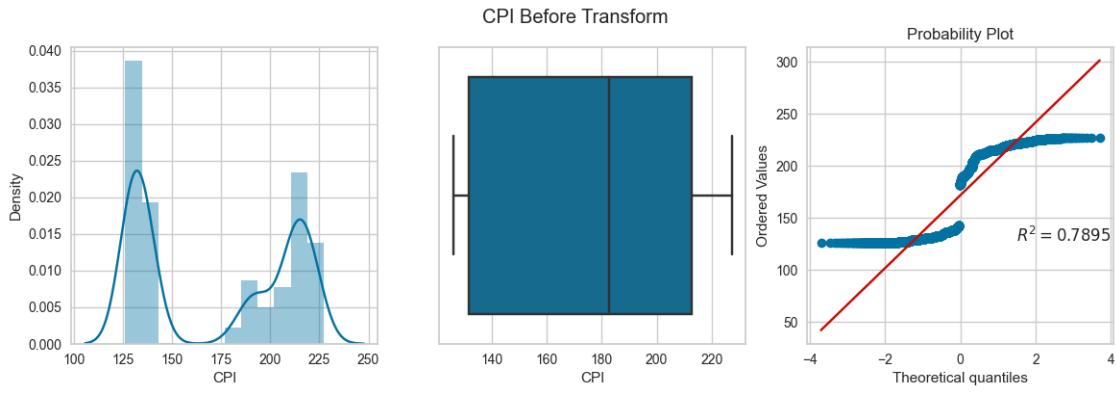


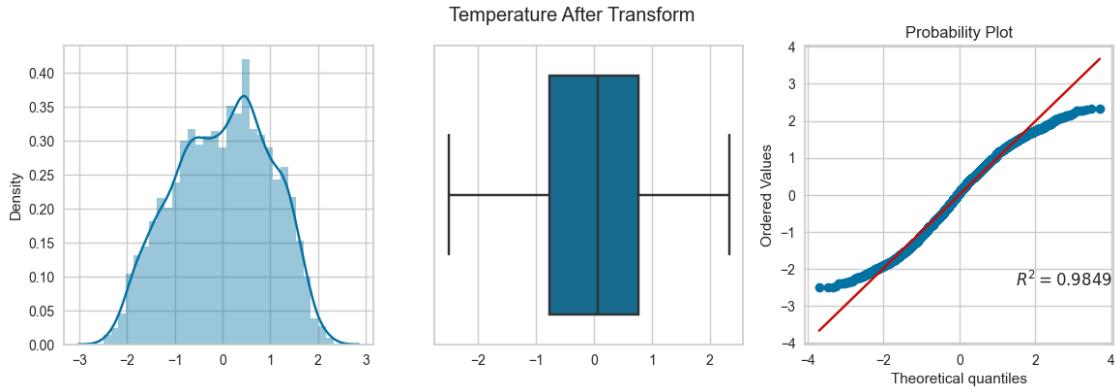
Fuel\_Price Before Transform



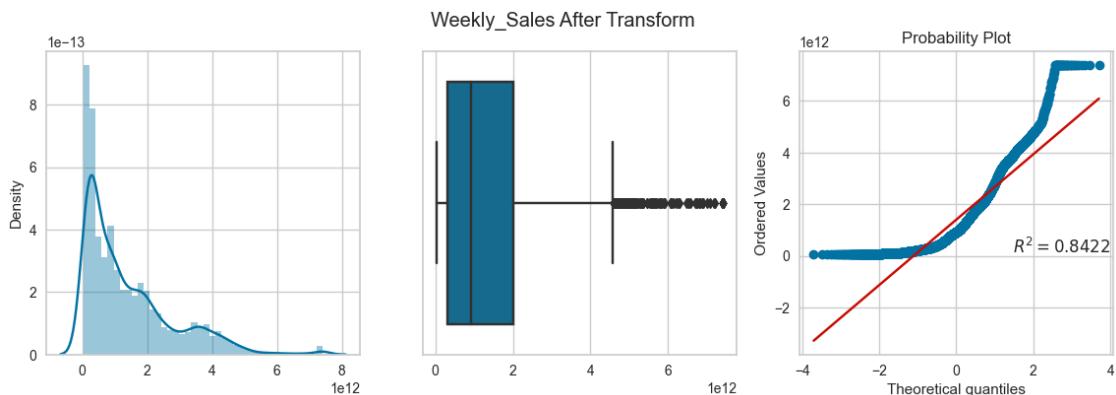
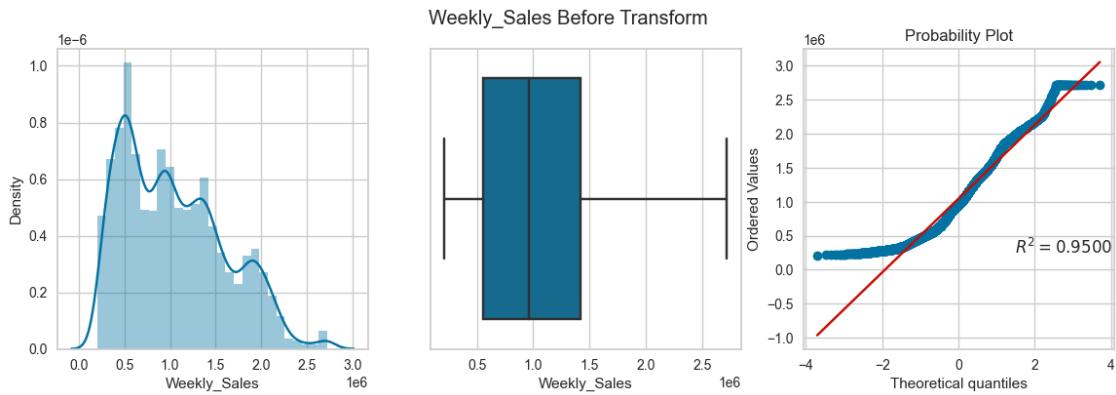
Fuel\_Price After Transform

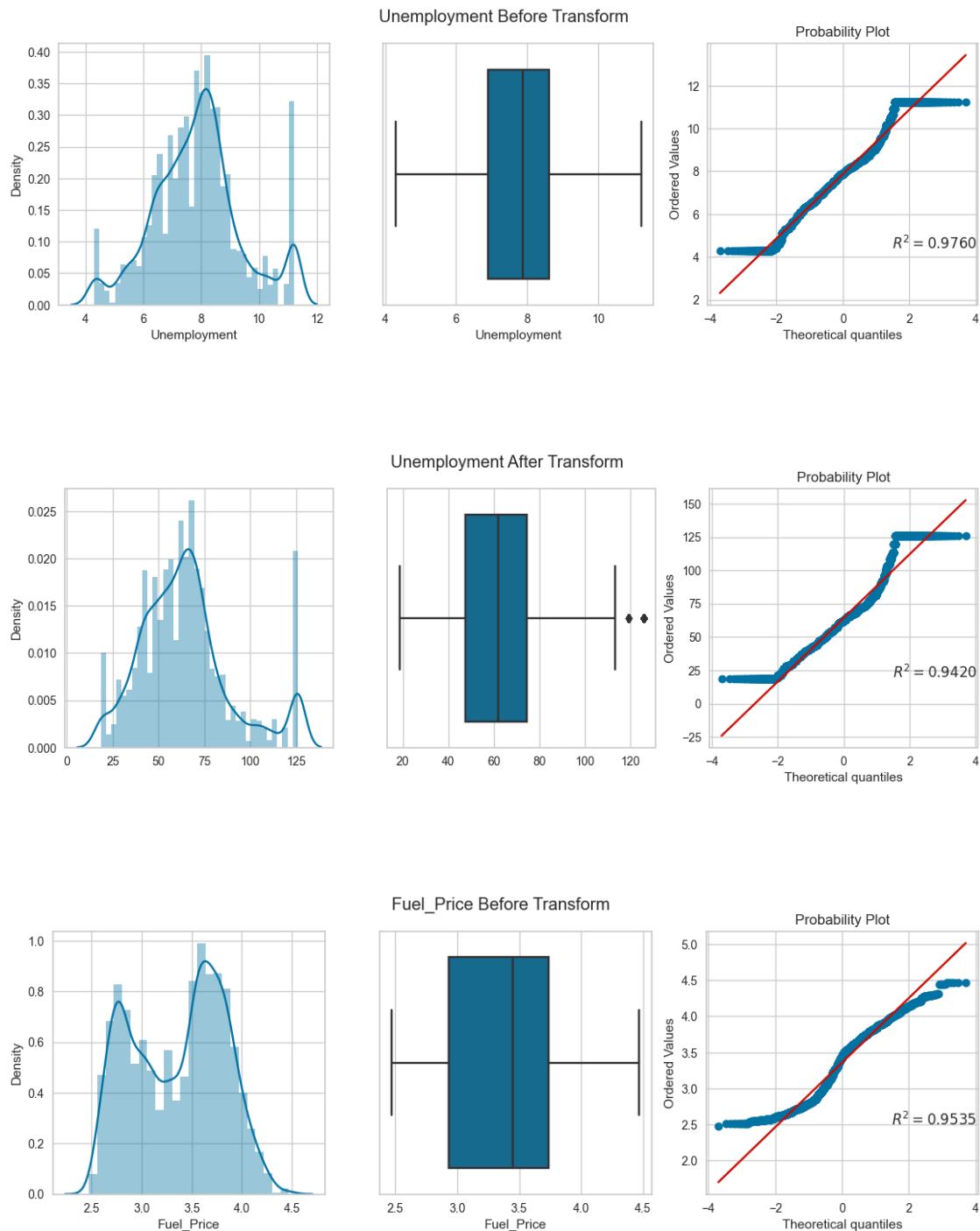


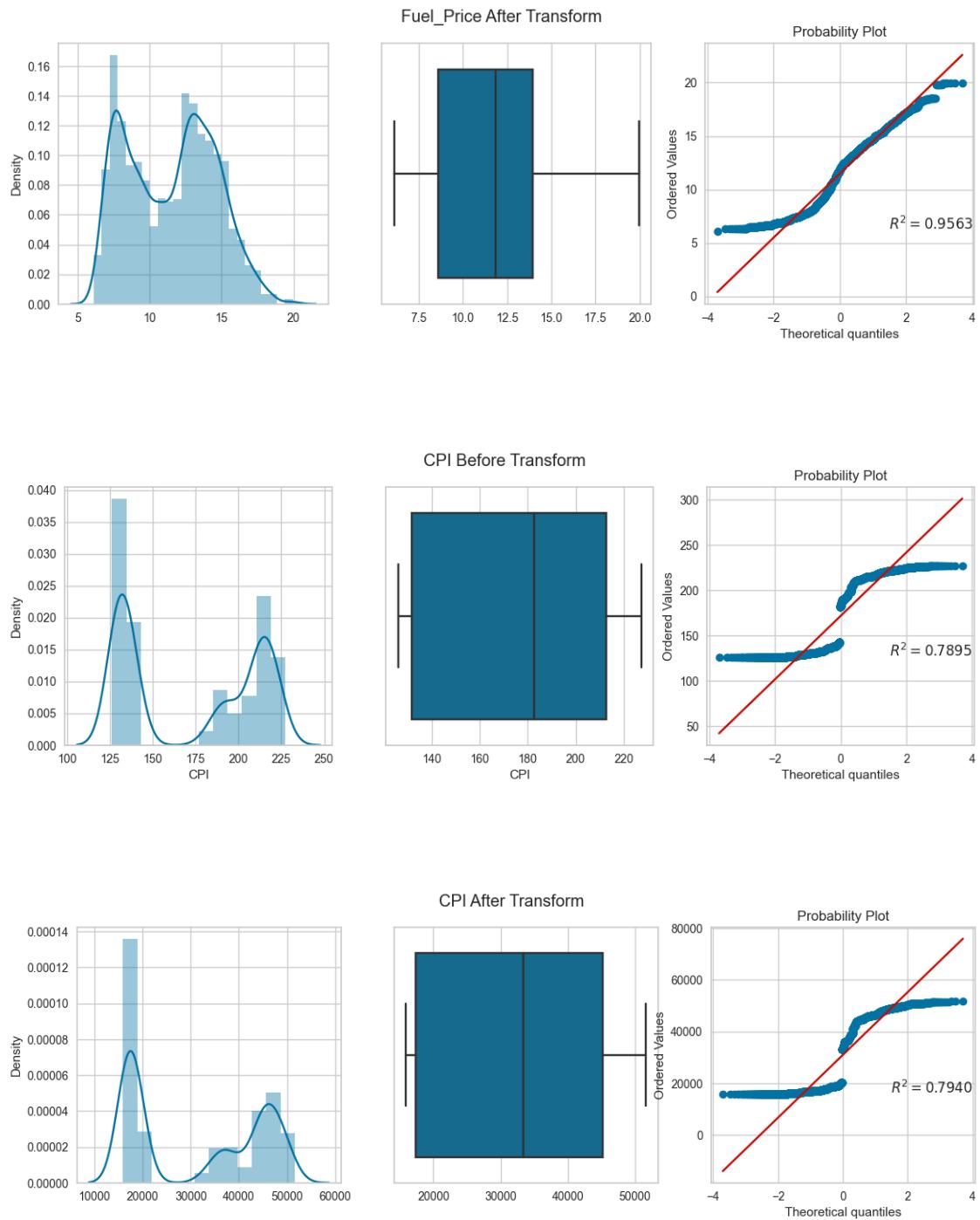


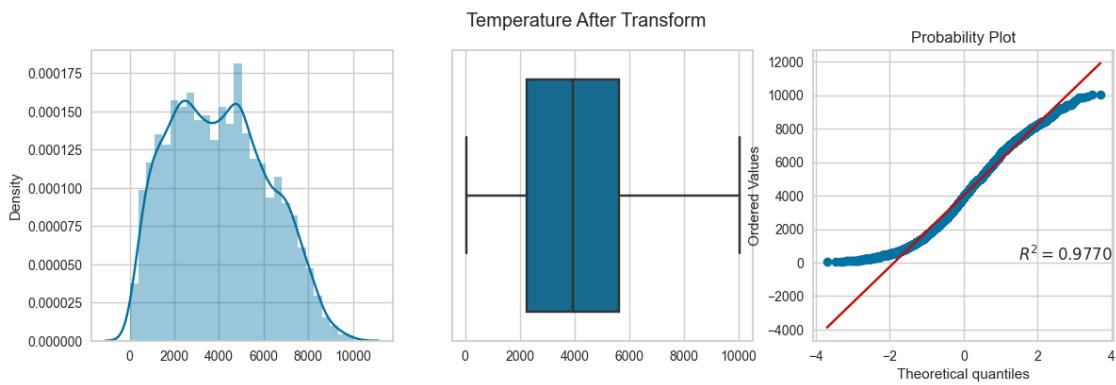
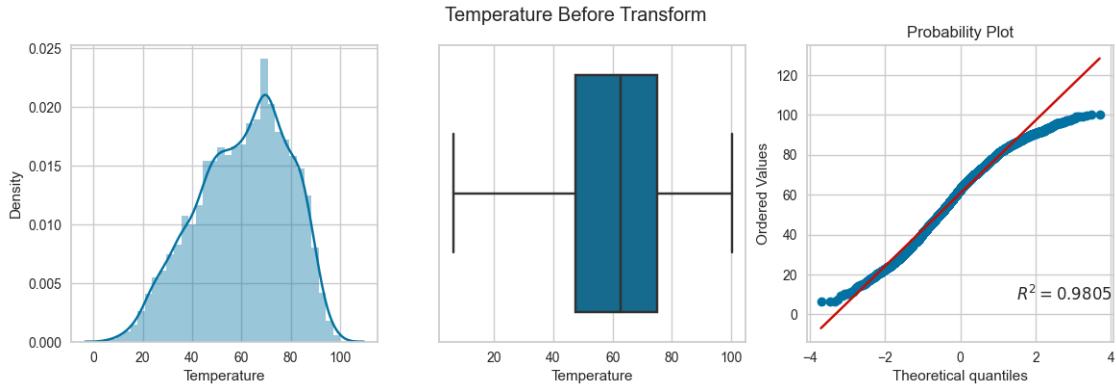


```
[28]: for col in skewed_cols:
    apply_transform(FunctionTransformer(lambda x: x**2), col)
```

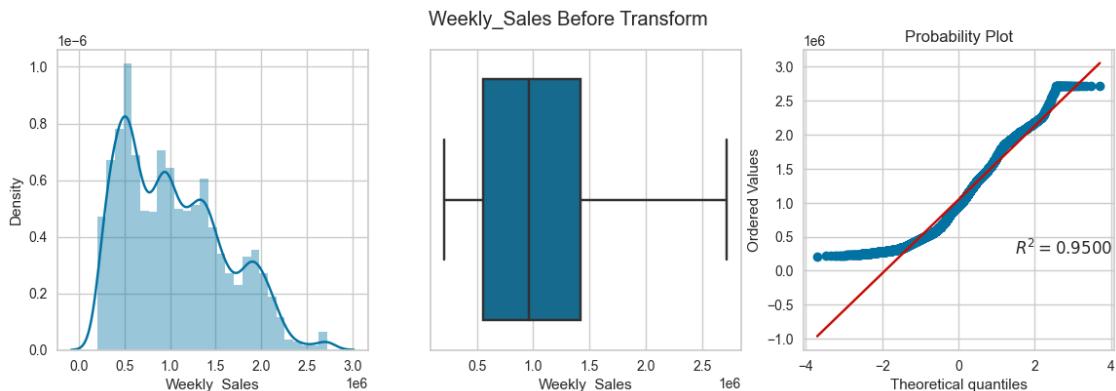


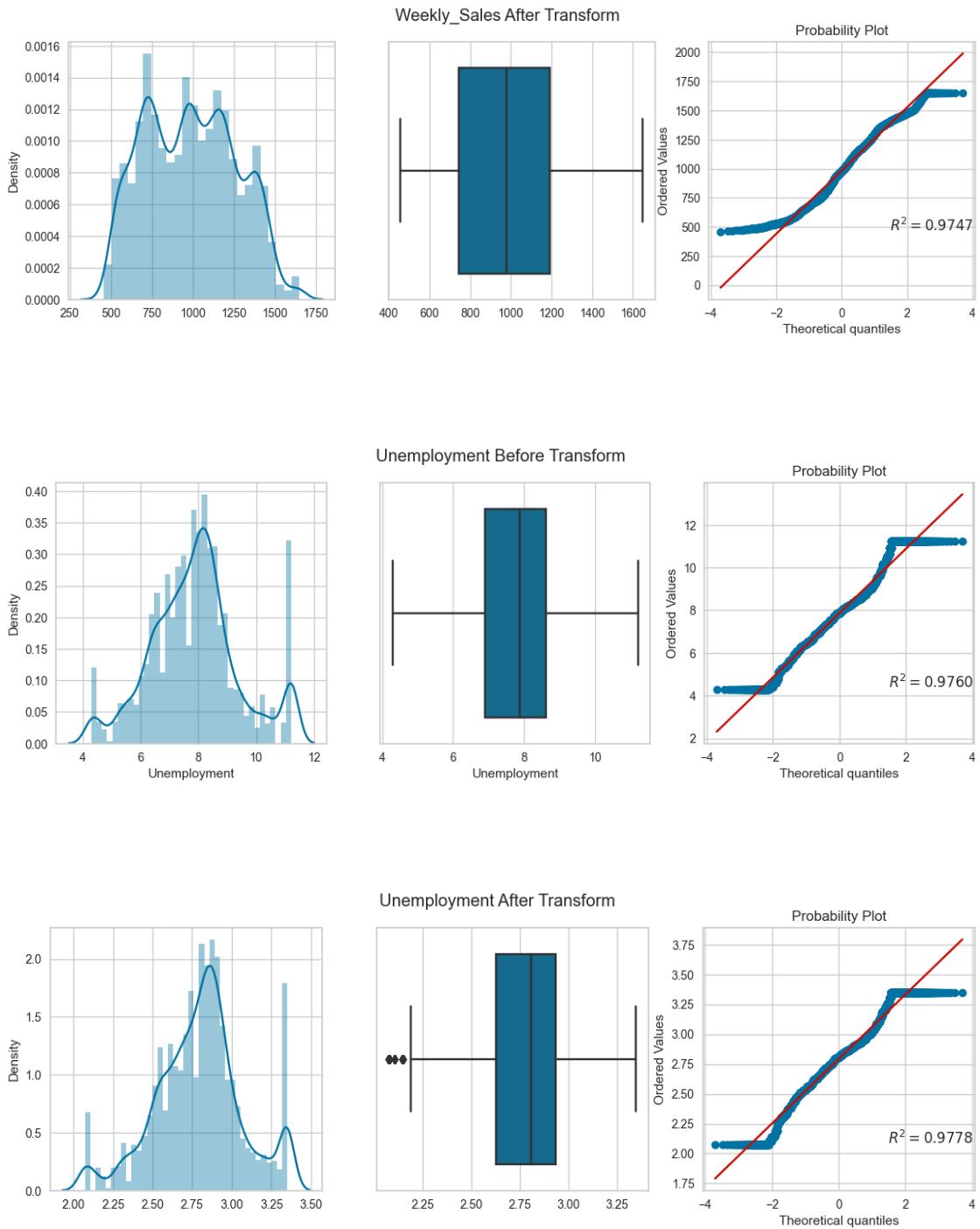


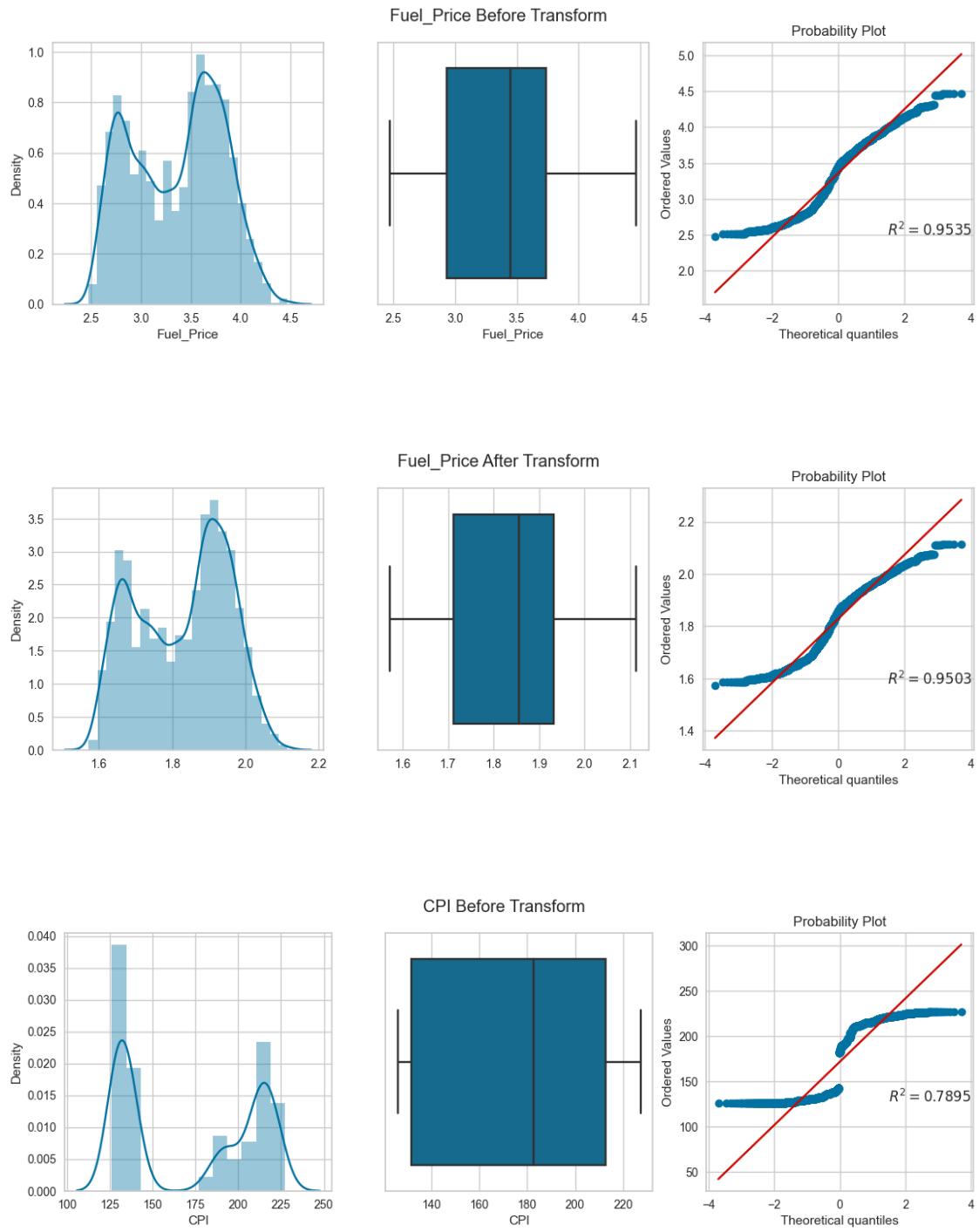


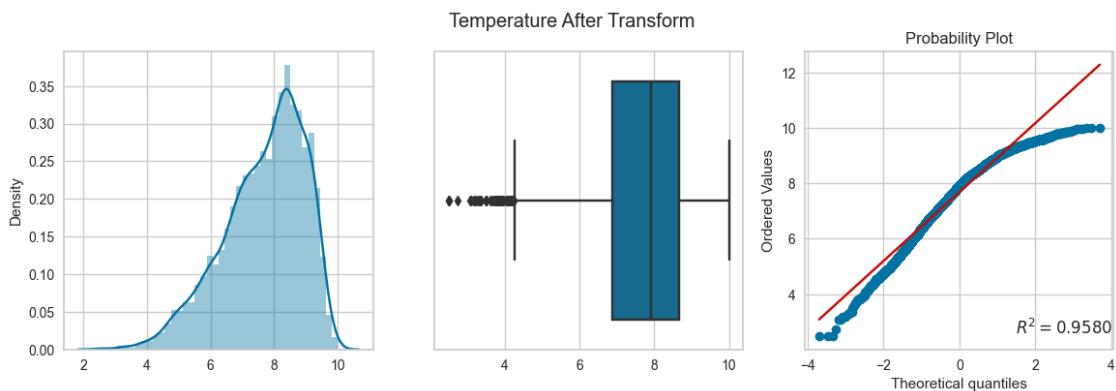
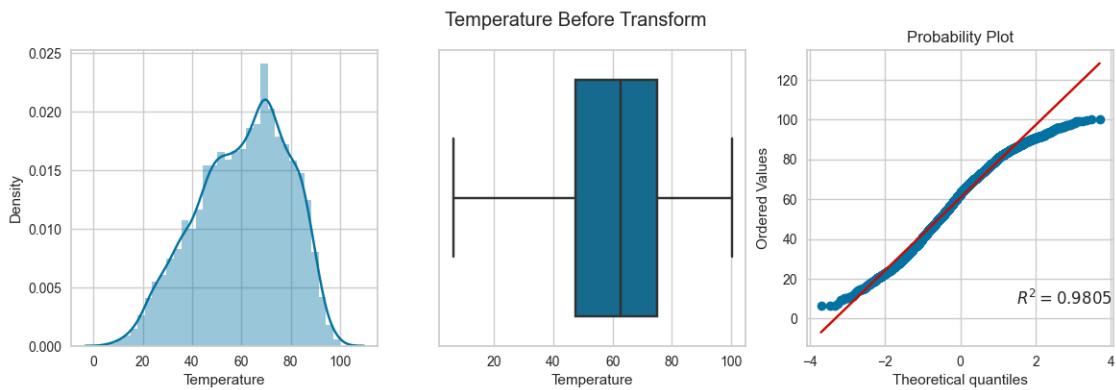
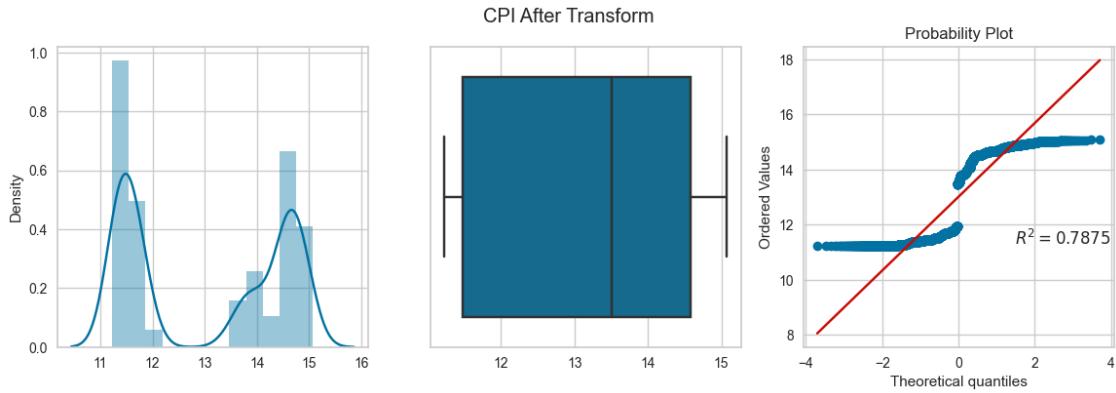


```
[29]: for col in skewed_cols:
    apply_transform(FunctionTransformer(np.sqrt), col)
```

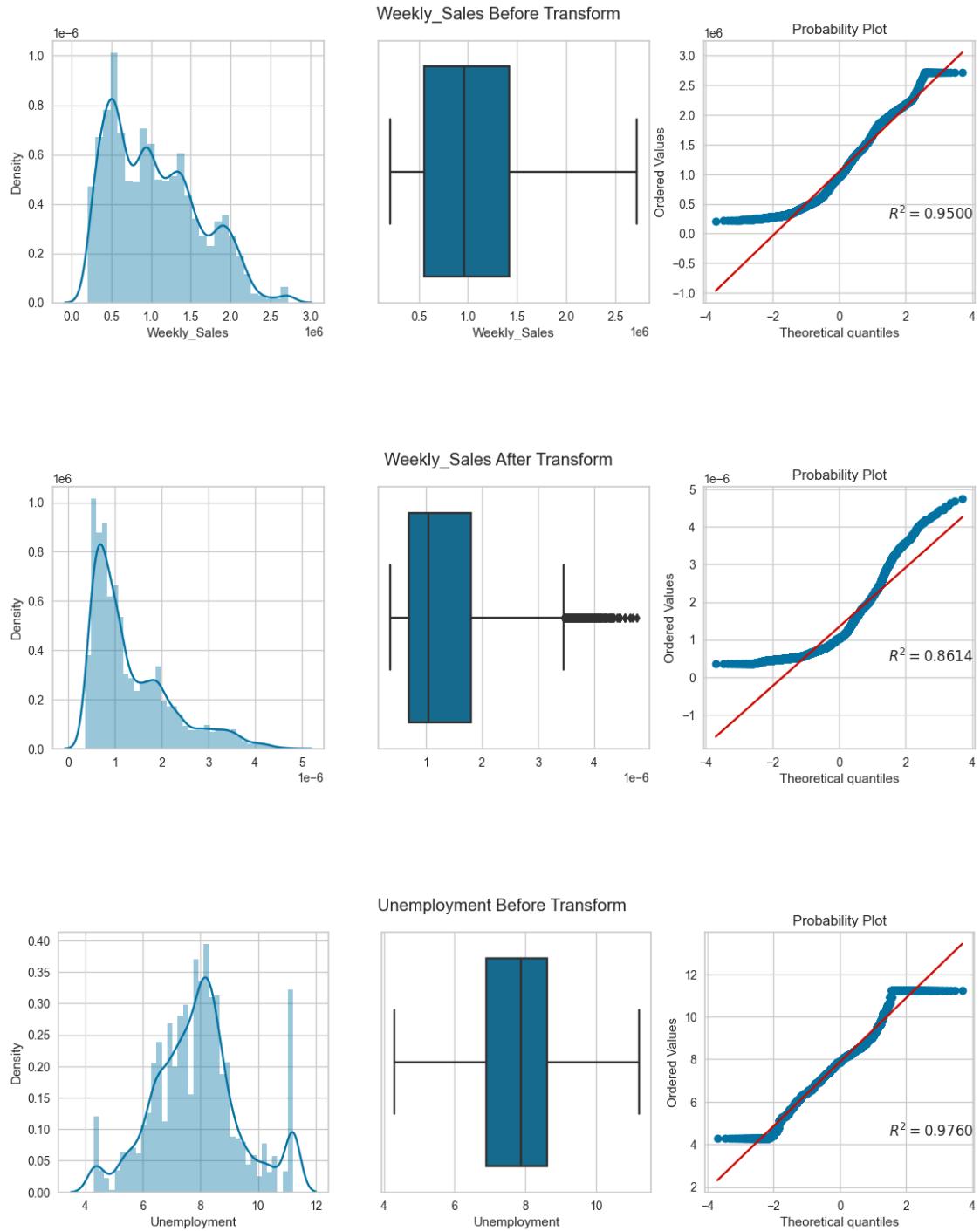


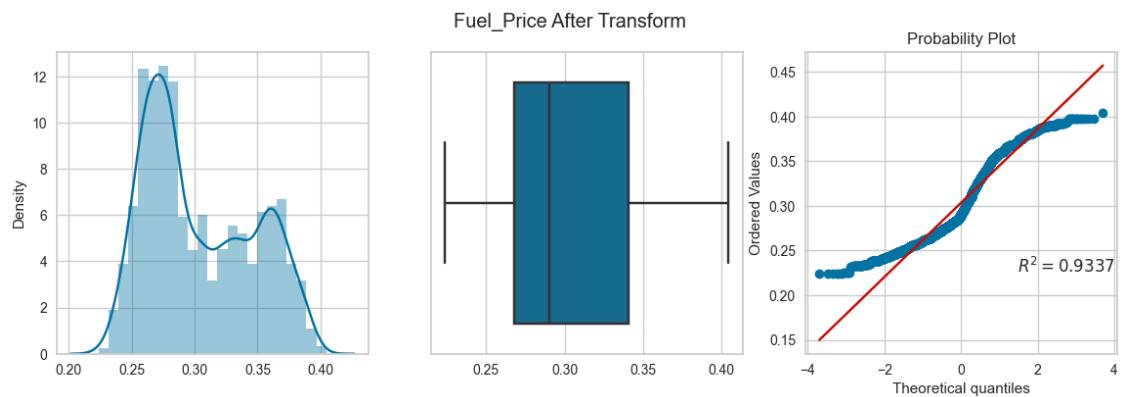
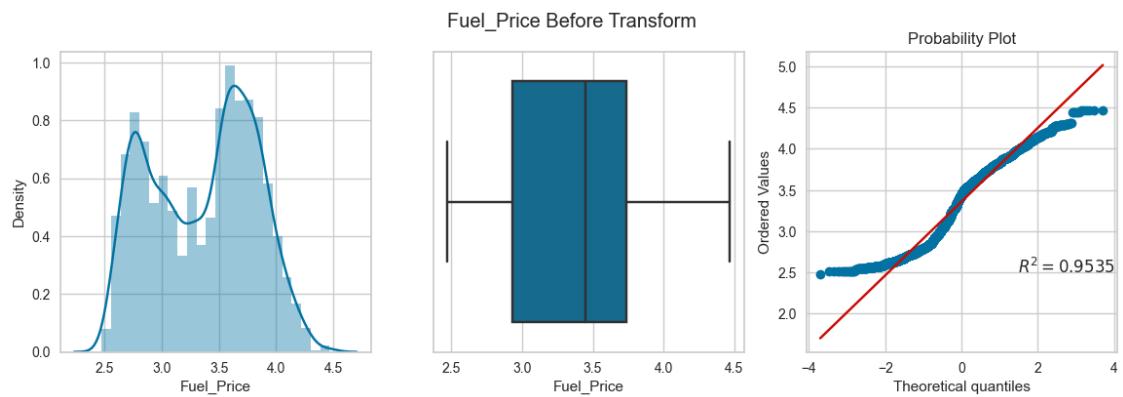
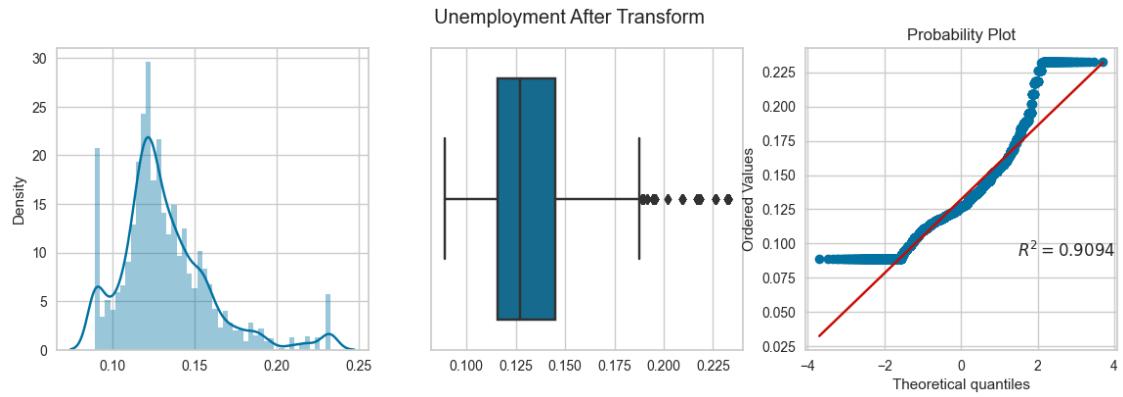


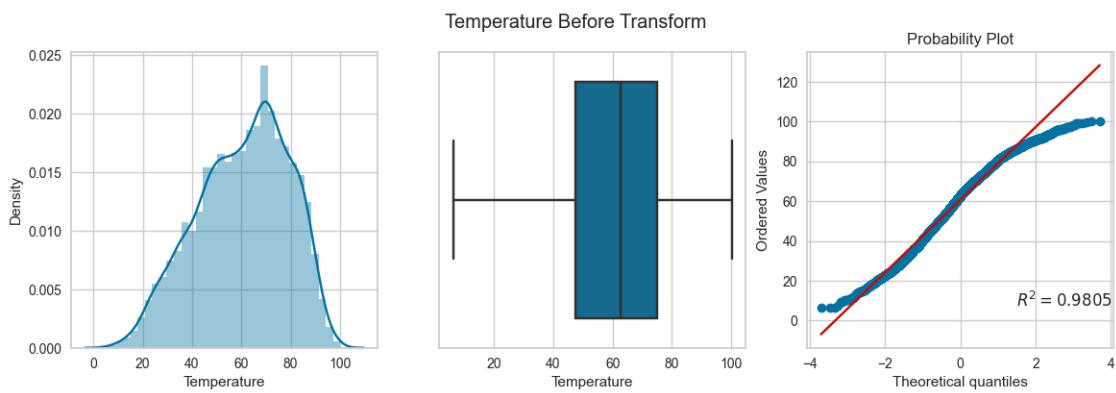
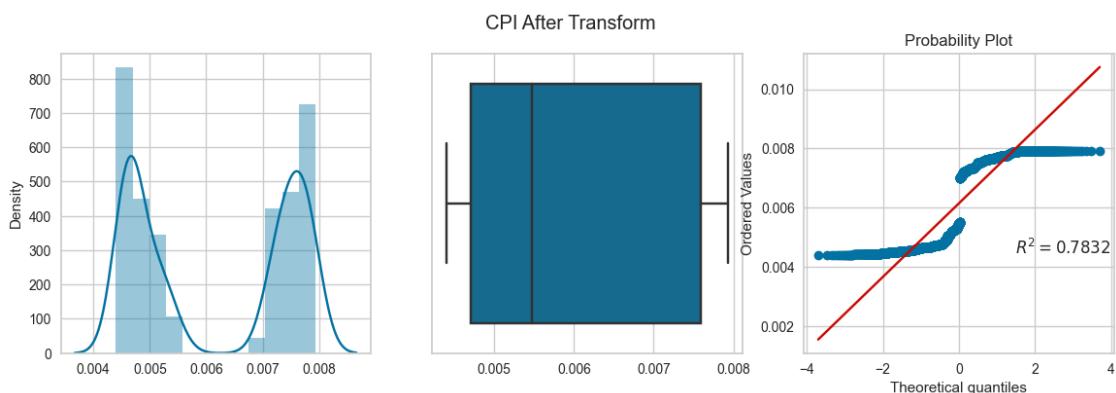
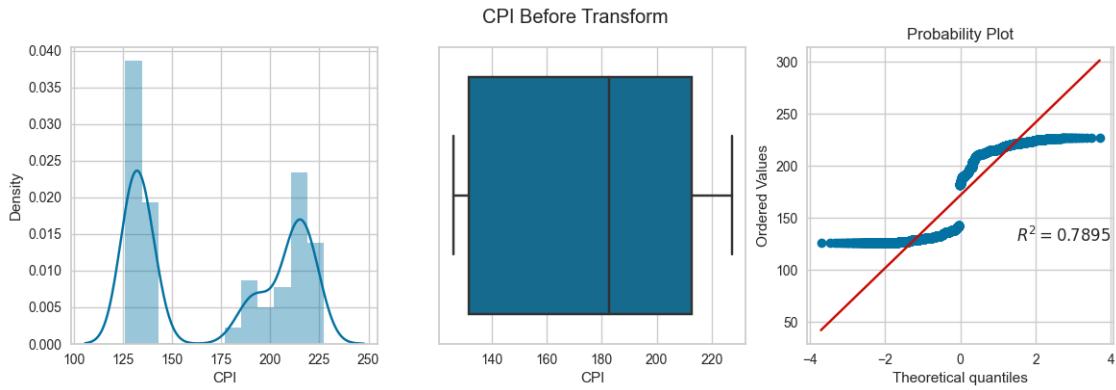


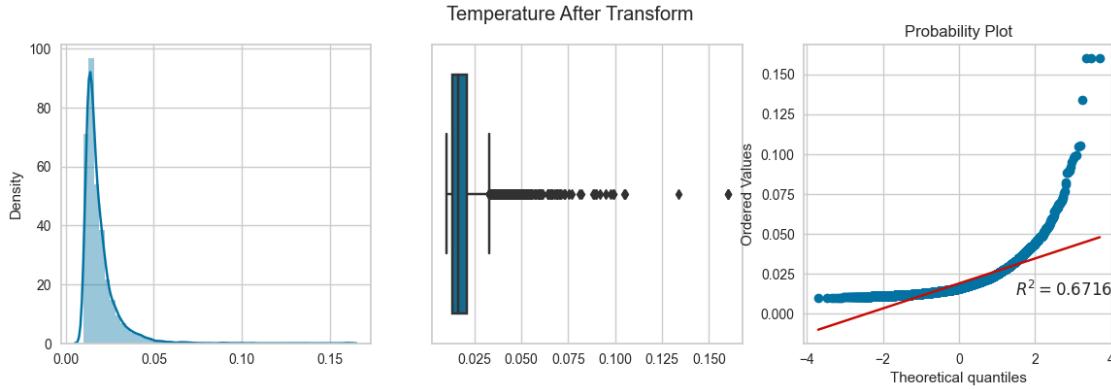


```
[30]: for col in skewed_cols:
    apply_transform(FunctionTransformer(lambda x: 1/(x+0.0000001)), col)
```









Log Transform: None  
Power Transform: Weekly\_Sales, Unemployment, Temperature Sqrt  
Transform: None  
Square Transform: Fuel\_Price, CPI  
Reciprocal Transform: None

```
[31]: pt = PowerTransformer()
pt_cols = ['Weekly_Sales', 'Unemployment', 'Temperature']

for col in pt_cols:
    col_tf = pt.fit_transform(df[[col]])
    col_tf = np.array(col_tf).reshape(col_tf.shape[0])
    df[col] = col_tf
```

```
[32]: ft = FunctionTransformer(lambda x: x**2)
ft_cols = ['Fuel_Price', 'CPI']

for col in ft_cols:
    col_tf = ft.fit_transform(df[[col]])
    col_tf = np.array(col_tf).reshape(col_tf.shape[0])
    df[col] = col_tf
```

### 0.5.3 Feature Extraction

```
[33]: df.Date = pd.to_datetime(df.Date, errors='coerce')
```

```
[34]: df['Year'] = df.Date.dt.year
df['Month'] = df.Date.dt.month_name()
df['Day'] = df.Date.dt.day_name()
df['Days_In_Month'] = df.Date.dt.daysinmonth
df['Is_Leap_Year'] = df.Date.dt.is_leap_year
df['Is_Quarter_Start'] = df.Date.dt.is_quarter_start
df['Is_Quarter_End'] = df.Date.dt.is_quarter_end
df['Is_Year_End'] = df.Date.dt.is_year_end
df['Is_Month_Start'] = df.Date.dt.is_month_start
df['Is_Month_End'] = df.Date.dt.is_month_end
```

```
df['Day_Of_Year'] = df.Date.dt.day_of_year  
df['Day_Of_Week'] = df.Date.dt.day_of_week
```

```
[35]: df.drop('Date',axis=1,inplace=True)
```

#### 0.5.4 Categorical Encoding

```
[36]: oe = OrdinalEncoder(categories=[[ 'January', 'February', 'March', 'April', 'May', 'June', 'July', 'August', 'September', 'October', 'November', 'December']])  
df.Month = oe.fit_transform(df[['Month']])  
df.Month = df.Month.astype(int)
```

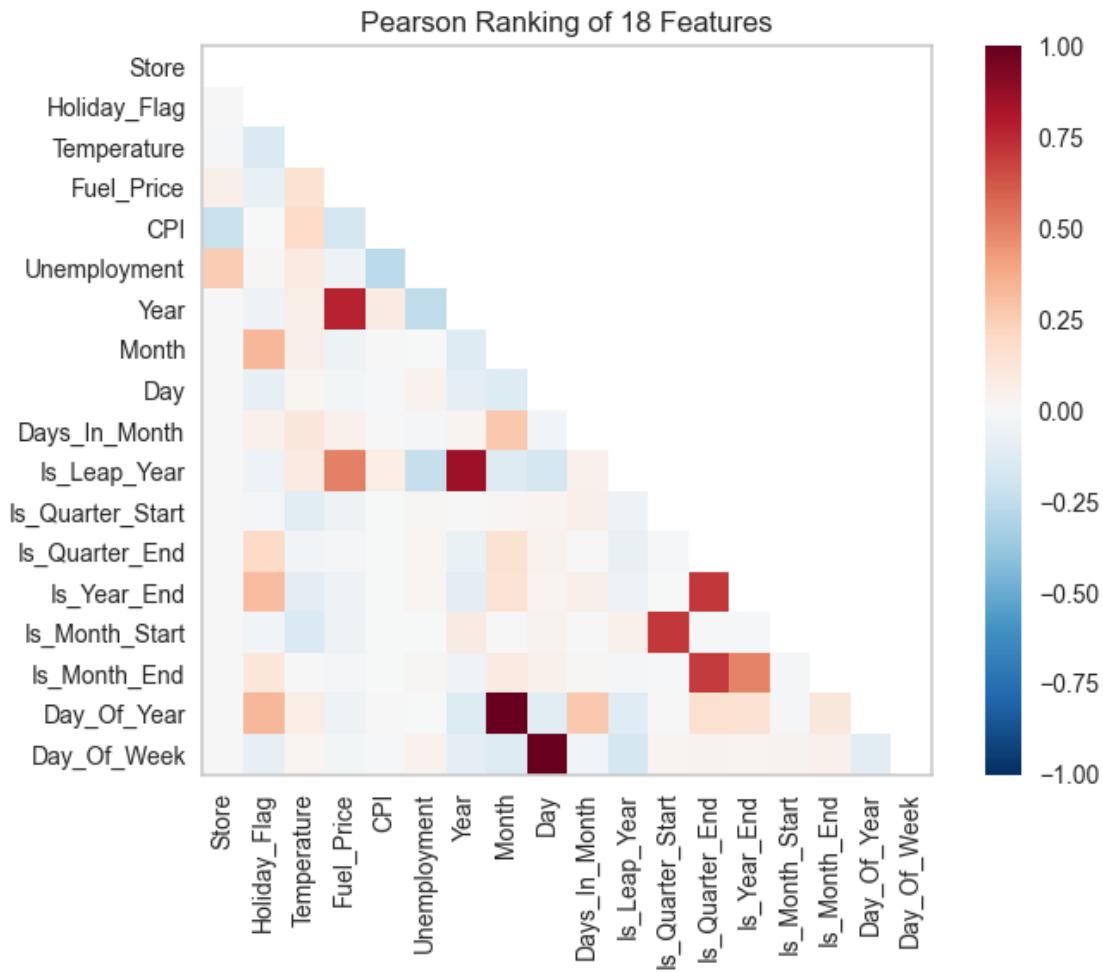
```
[37]: oe = OrdinalEncoder(categories=[[ 'Monday', 'Tuesday', 'Wednesday', 'Thursday', 'Friday', 'Saturday', 'Sunday']])  
df.Day = oe.fit_transform(df[['Day']])  
df.Day = df.Day.astype(int)
```

```
[38]: df.Is_Leap_Year = df.Is_Leap_Year.astype(int)  
df.Is_Quarter_Start = df.Is_Quarter_Start.astype(int)  
df.Is_Quarter_End = df.Is_Quarter_End.astype(int)  
df.Is_Year_End = df.Is_Year_End.astype(int)  
df.Is_Month_Start = df.Is_Month_Start.astype(int)  
df.Is_Month_End = df.Is_Month_End.astype(int)
```

#### 0.5.5 Feature Splitting

```
[39]: X = df.drop('Weekly_Sales',axis=1)  
y = df.Weekly_Sales
```

```
[40]: ranking_2d = Rank2D(algorithm='pearson')  
ranking_2d.fit(X,y)  
ranking_2d.transform(X)  
ranking_2d.show();
```



```
[41]: X_train, X_test, y_train, y_test = train_test_split(X,y,test_size=0.
    ↵3,random_state=101,shuffle=True)
```

## 0.5.6 Feature Selection

```
[42]: kbest = SelectKBest(k=6,score_func=f_regression)
kbest.fit(X_train,y_train)
```

```
[42]: SelectKBest(k=6, score_func=<function f_regression at 0x000001BE01581090>)
```

```
[43]: selected_features = kbest.get_feature_names_out()
selected_features
```

```
[43]: array(['Store', 'Temperature', 'CPI', 'Unemployment', 'Month',
       'Day_Of_Year'], dtype=object)
```

```
[44]: percentile = SelectPercentile(percentile=30,score_func=f_regression)
percentile.fit(X_train,y_train)
```

```
[44]: SelectPercentile(percentile=30,
score_func=<function f_regression at 0x000001BE01581090>)
```

```
[45]: selected_features = percentile.get_feature_names_out()
selected_features
```

```
[45]: array(['Store', 'Temperature', 'CPI', 'Unemployment', 'Month',
'Day_Of_Year'], dtype=object)
```

```
[46]: sfm = SelectFromModel(RandomForestRegressor(),threshold='median')
sfm.fit(X_train,y_train)
```

```
[46]: SelectFromModel(estimator=RandomForestRegressor(), threshold='median')
```

```
[47]: selected_features = sfm.get_feature_names_out()
selected_features
```

```
[47]: array(['Store', 'Holiday_Flag', 'Temperature', 'Fuel_Price', 'CPI',
'Unemployment', 'Month', 'Day_Of_Year', 'Day_Of_Week'],
dtype=object)
```

```
[48]: rfe =_
    RFECV(estimator=RandomForestRegressor(),min_features_to_select=6,step=2,cv=5,verbose=1)
rfe.fit(X_train,y_train)
```

```
Fitting estimator with 18 features.
Fitting estimator with 16 features.
Fitting estimator with 14 features.
Fitting estimator with 12 features.
Fitting estimator with 10 features.
Fitting estimator with 8 features.
Fitting estimator with 18 features.
Fitting estimator with 16 features.
Fitting estimator with 14 features.
Fitting estimator with 12 features.
Fitting estimator with 10 features.
Fitting estimator with 8 features.
Fitting estimator with 18 features.
Fitting estimator with 16 features.
Fitting estimator with 14 features.
Fitting estimator with 12 features.
Fitting estimator with 10 features.
Fitting estimator with 8 features.
Fitting estimator with 18 features.
Fitting estimator with 16 features.
```

```
Fitting estimator with 14 features.  
Fitting estimator with 12 features.  
Fitting estimator with 10 features.  
Fitting estimator with 8 features.  
Fitting estimator with 18 features.  
Fitting estimator with 16 features.  
Fitting estimator with 14 features.  
Fitting estimator with 12 features.  
Fitting estimator with 10 features.  
Fitting estimator with 8 features.  
Fitting estimator with 18 features.  
Fitting estimator with 16 features.  
Fitting estimator with 14 features.
```

```
[48]: RFECV(cv=5, estimator=RandomForestRegressor(), min_features_to_select=6, step=2,  
          verbose=1)
```

```
[49]: selected_features = rfe.get_feature_names_out()  
selected_features
```

```
[49]: array(['Store', 'Holiday_Flag', 'Temperature', 'Fuel_Price', 'CPI',  
           'Unemployment', 'Year', 'Month', 'Day', 'Days_In_Month',  
           'Day_Of_Year', 'Day_Of_Week'], dtype=object)
```

```
[50]: sfs =  
      SequentialFeatureSelector(estimator=RandomForestRegressor(), n_features_to_select=6, direction=  
      sfs.fit(X_train,y_train)
```

```
[50]: SequentialFeatureSelector(cv=3, estimator=RandomForestRegressor(),  
                               n_features_to_select=6)
```

```
[51]: selected_features = sfs.get_feature_names_out()  
selected_features
```

```
[51]: array(['Store', 'Holiday_Flag', 'CPI', 'Unemployment', 'Day',  
           'Day_Of_Year'], dtype=object)
```

```
[52]: lasso = SelectFromModel(estimator=LassoCV(), threshold='0.95*mean')  
lasso.fit(X_train,y_train)
```

```
[52]: SelectFromModel(estimator=LassoCV(), threshold='0.95*mean')
```

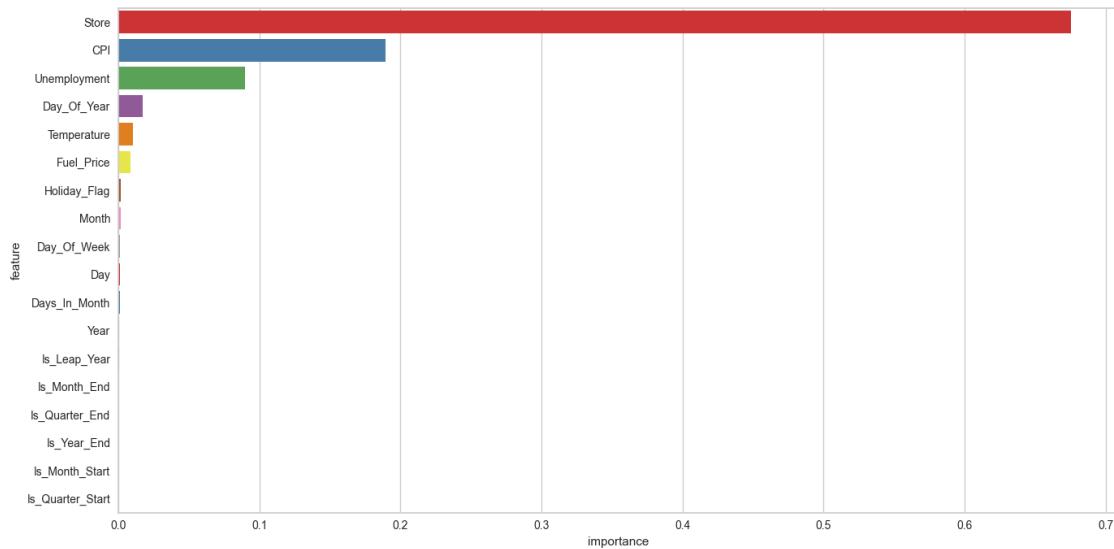
```
[53]: selected_features = lasso.get_feature_names_out()  
selected_features
```

```
[53]: array(['Store'], dtype=object)
```

```
[54]: rf = RandomForestRegressor()
rf.fit(X_train,y_train)
```

```
[54]: RandomForestRegressor()
```

```
[55]: feat_imps = pd.DataFrame({'feature': X_train.columns, 'importance': rf.
    ↪feature_importances_}).sort_values('importance',ascending=False)
plt.figure(figsize=(14,7))
sns.
    ↪barplot(x='importance',y='feature',data=feat_imps,orient='horizontal',palette='Set1')
plt.tight_layout();
```



```
[56]: df.corr()['Weekly_Sales'].sort_values(ascending=False)[1:]
```

```
[56]: Month          0.050730
Day_Of_Year      0.048769
Holiday_Flag     0.029496
Fuel_Price       0.016294
Is_Leap_Year     -0.006772
Days_In_Month   -0.008107
Year             -0.009913
Is_Month_Start  -0.011720
Is_Quarter_Start -0.011894
Is_Year_End      -0.020678
Day              -0.020999
Day_Of_Week      -0.020999
Is_Month_End     -0.022571
Is_Quarter_End   -0.025321
```

```

CPI           -0.082511
Temperature   -0.085931
Unemployment -0.087557
Store          -0.308973
Name: Weekly_Sales, dtype: float64

```

Final selected features: Store, CPI, Unemployment, Day\_Of\_Year, Temperature, Month, Holiday\_Flag

```
[57]: selected_features = [
    'Store', 'CPI', 'Unemployment', 'Day_Of_Year', 'Temperature', 'Month', 'Holiday_Flag']

X_train = X_train[selected_features]
X_test = X_test[selected_features]
```

### 0.5.7 Feature Scaling

```
[58]: scaler = StandardScaler()
features = X_train.columns
X_train = scaler.fit_transform(X_train)
X_train = pd.DataFrame(X_train, columns=features)
X_test = scaler.transform(X_test)
X_test = pd.DataFrame(X_test, columns=features)
X_train.head()
```

```
[58]:      Store      CPI  Unemployment  Day_Of_Year  Temperature  Month \
0  1.304226 -0.974622     -1.870001   -0.865484   -2.253026 -0.728108
1  1.458359 -1.013378     -0.580989    1.183872    0.511933  1.078226
2 -1.470166 -1.054563     -1.270860   -0.234913    0.109179 -0.125997
3 -1.470166 -1.109167     -0.292376   -1.397529    0.695575 -1.330220
4 -1.470166 -1.039096     -1.848716    0.740502   -1.509232  0.777170

      Holiday_Flag
0      -0.268562
1      -0.268562
2      -0.268562
3      -0.268562
4      -0.268562
```

## 0.6 Model Training & Evaluation

```
[59]: models = []
mape_scores = []
rmse_scores = []
r2_scores = []
```

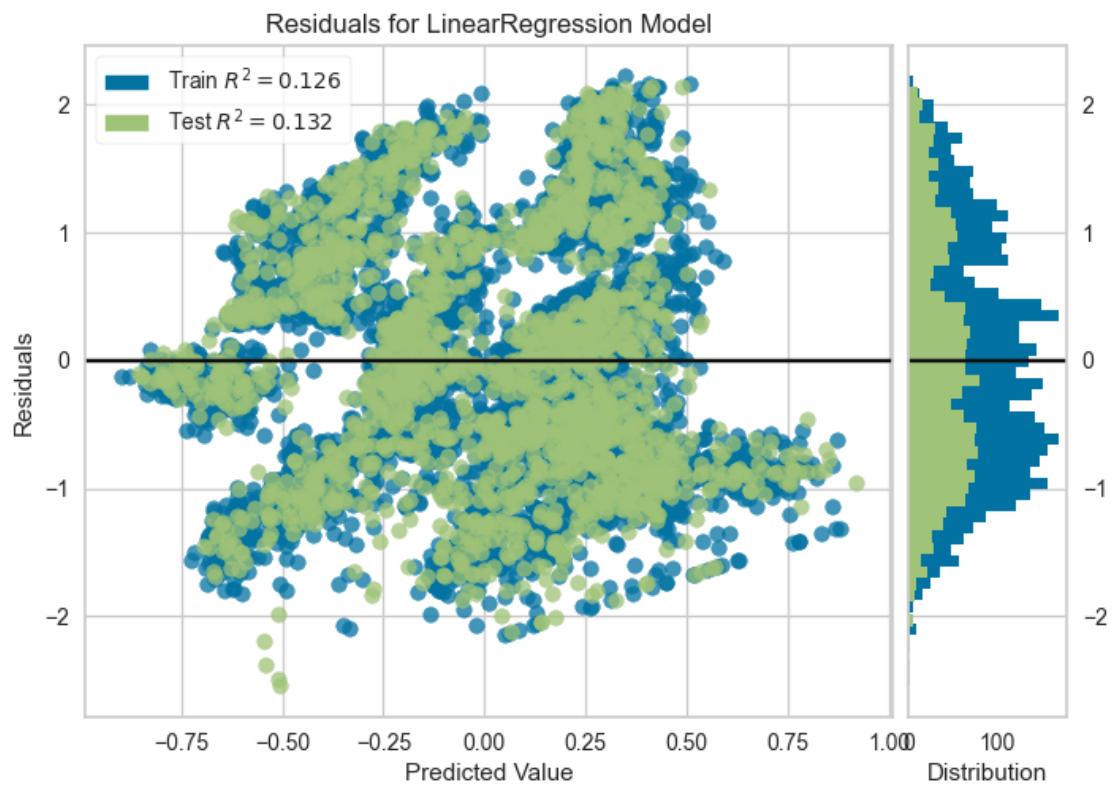
```
[144]: def train_and_evaluate_model(model):
    model.fit(X_train,y_train)
    pred = model.predict(X_test)
    mape = mean_absolute_percentage_error(y_test,pred)
    print("Mean Absolute Percentage Error(MAPE) : ",mape)
    mape_scores.append(mape)
    rmse = np.sqrt(mean_squared_error(y_test,pred))
    print("Root Mean Squared Error(RMSE) : ",rmse)
    rmse_scores.append(rmse)
    r2 = r2_score(y_test,pred)
    print("R2 Score: ",r2)
    r2_scores.append(r2)
    models.append(model)

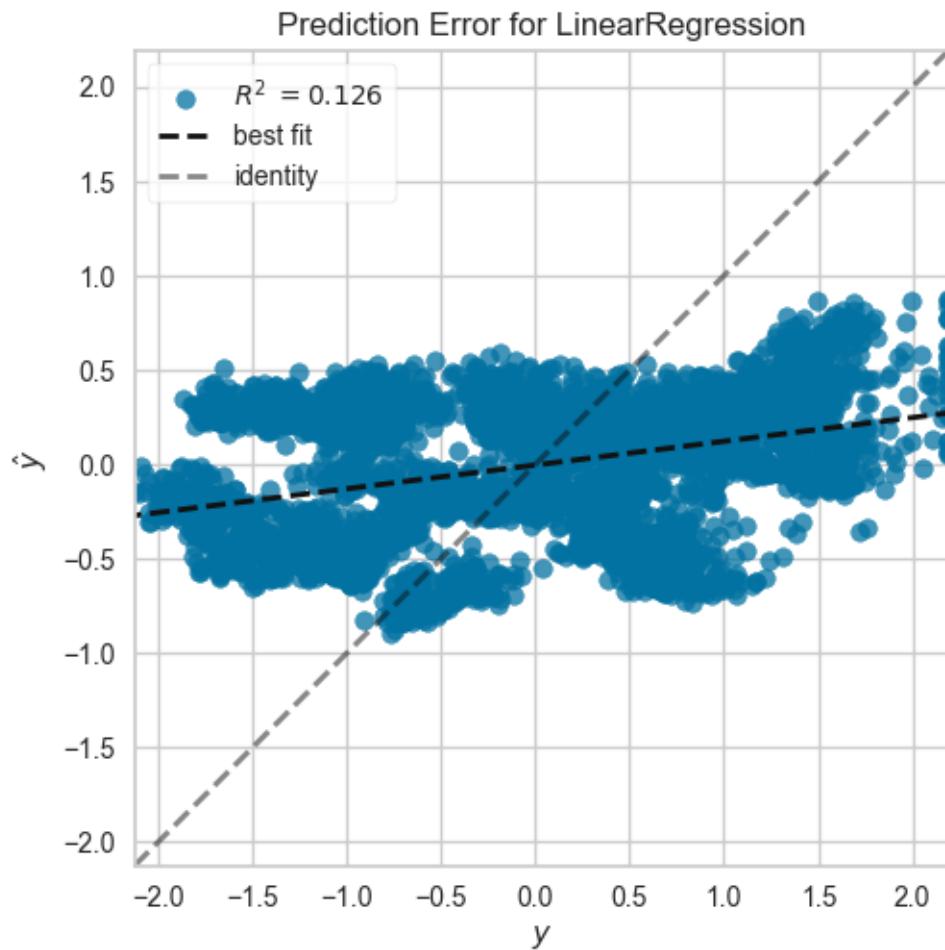
    if re.search('catboost',str(model)) == None or re.
    ↪search('Voting',str(model)) != None:
        residuals_plot = ResidualsPlot(model)
        residuals_plot.fit(X_train,y_train)
        residuals_plot.score(X_test,y_test)
        residuals_plot.show()
        predictions_error = PredictionError(model)
        predictions_error.score(X_train,y_train)
        predictions_error.fit(X_train,y_train)
        predictions_error.show()

    if str(model) == 'LassoCV()':
        visualizer = AlphaSelection(LassoCV(alphas=np.logspace(-10,10,400)))
        visualizer.fit(X_train,y_train)
        visualizer.show()
    elif str(model) == 'RidgeCV()':
        visualizer = AlphaSelection(RidgeCV(alphas=np.logspace(-10,10,400)))
        visualizer.fit(X_train,y_train)
        visualizer.show()
    elif str(model) == 'ElasticNetCV()':
        visualizer = AlphaSelection(ElasticNetCV(alphas=np.
        ↪logspace(-10,10,400)))
        visualizer.fit(X_train,y_train)
        visualizer.show()
```

```
[61]: train_and_evaluate_model(LinearRegression())
```

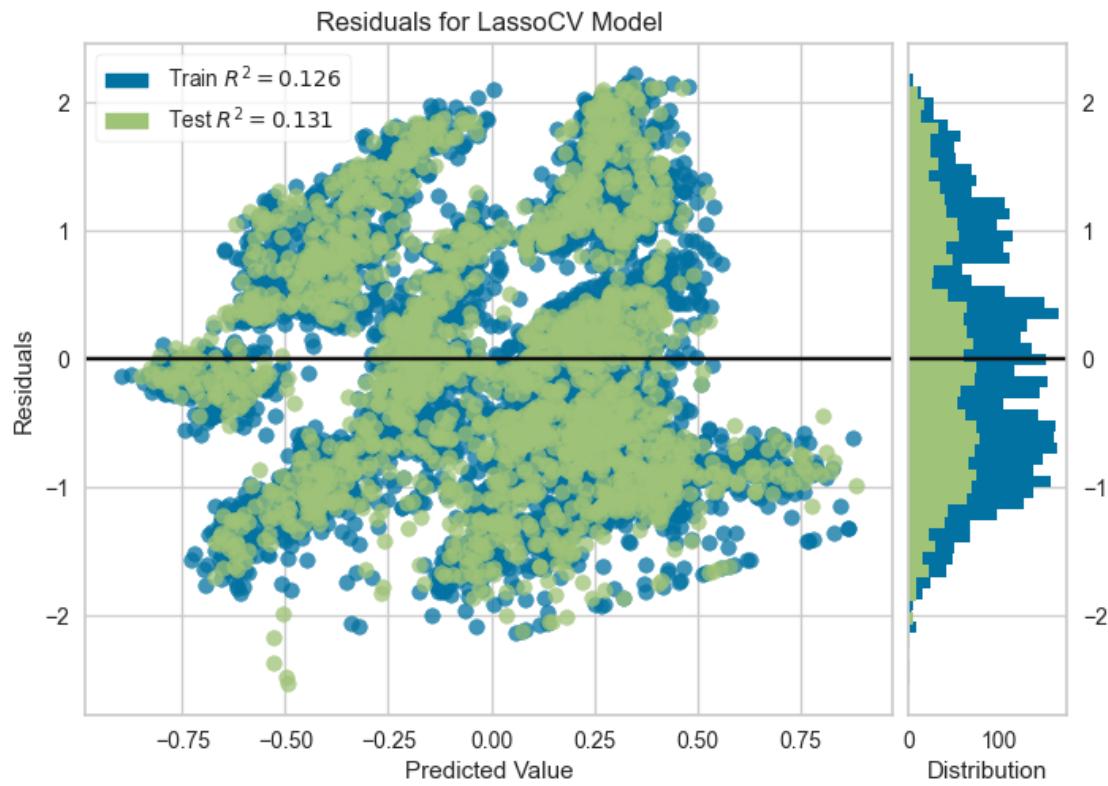
```
Mean Absolute Percentage Error(MAPE): 1.8588602855362242
Root Mean Squared Error(RMSE): 0.9428056032781673
R2 Score: 0.13166572867067572
```

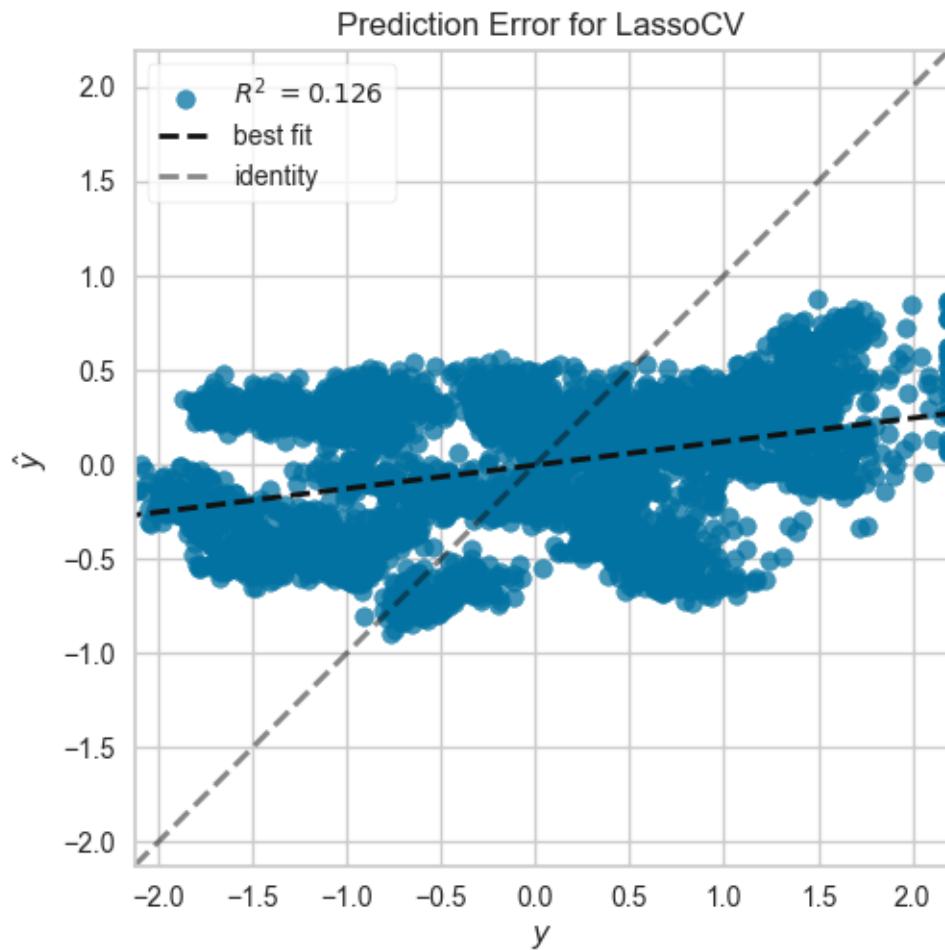


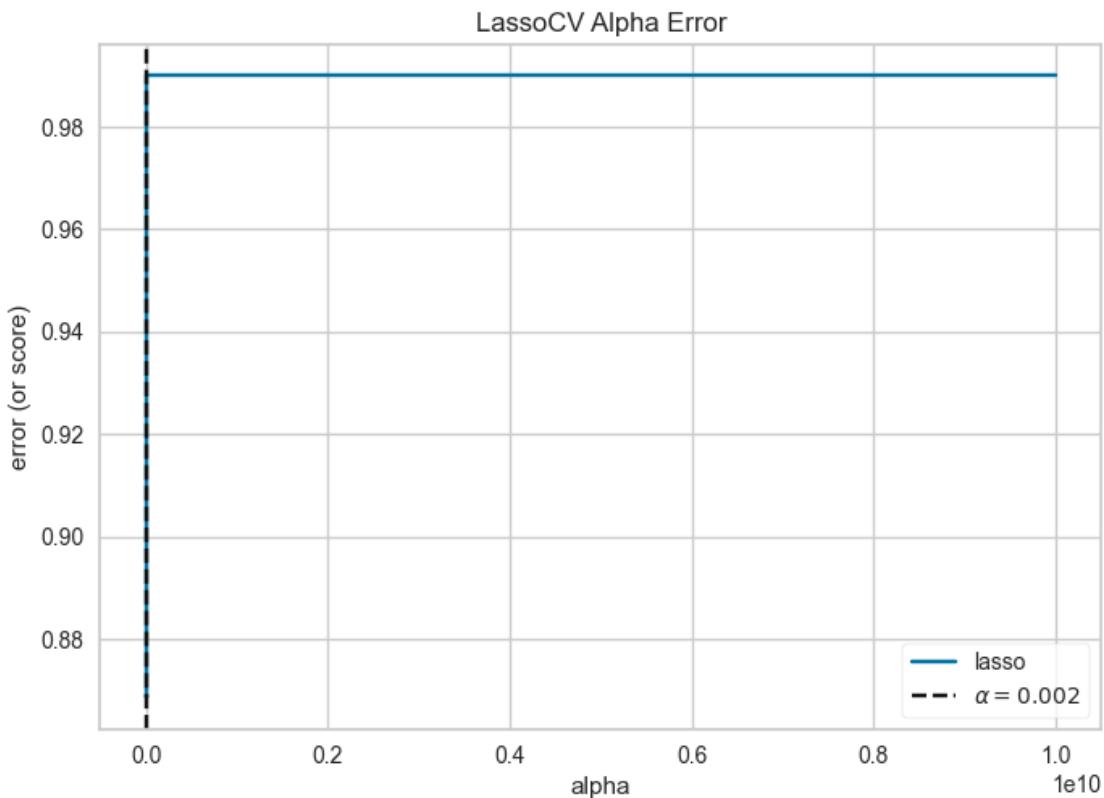


```
[62]: train_and_evaluate_model(LassoCV())
```

```
Mean Absolute Percentage Error(MAPE): 1.87217359396011
Root Mean Squared Error(RMSE): 0.9430963176756088
R2 Score: 0.131130143830821
```

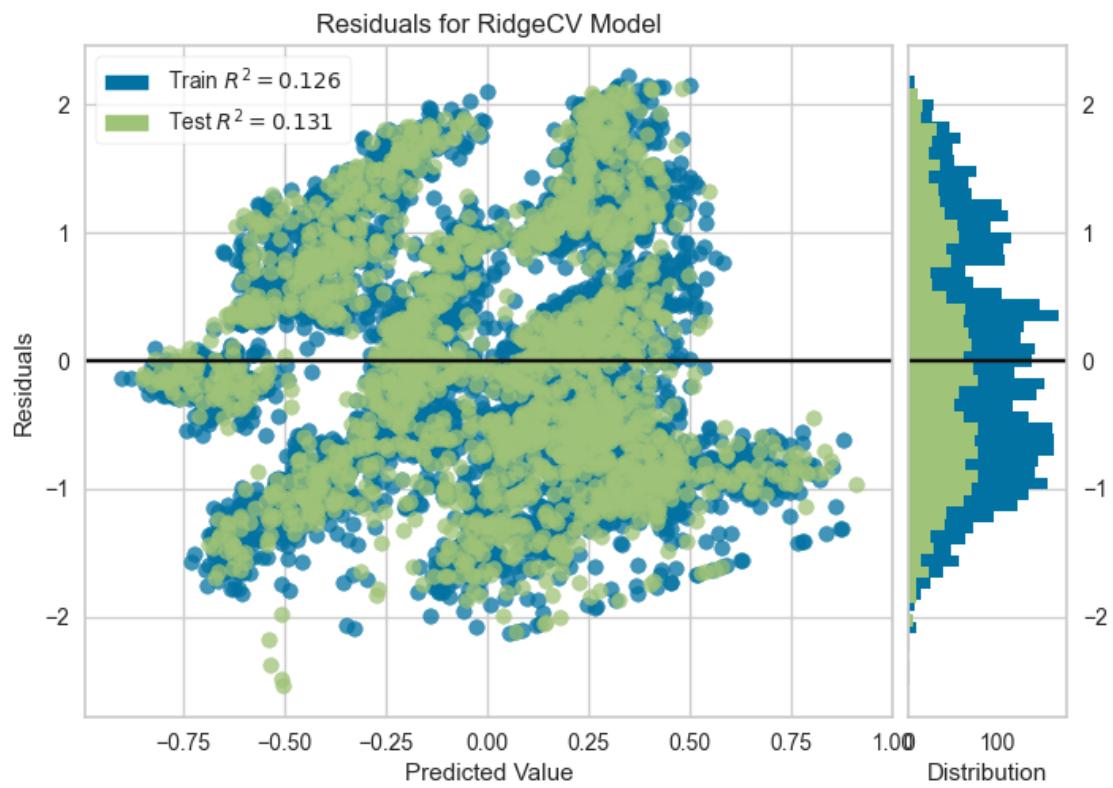


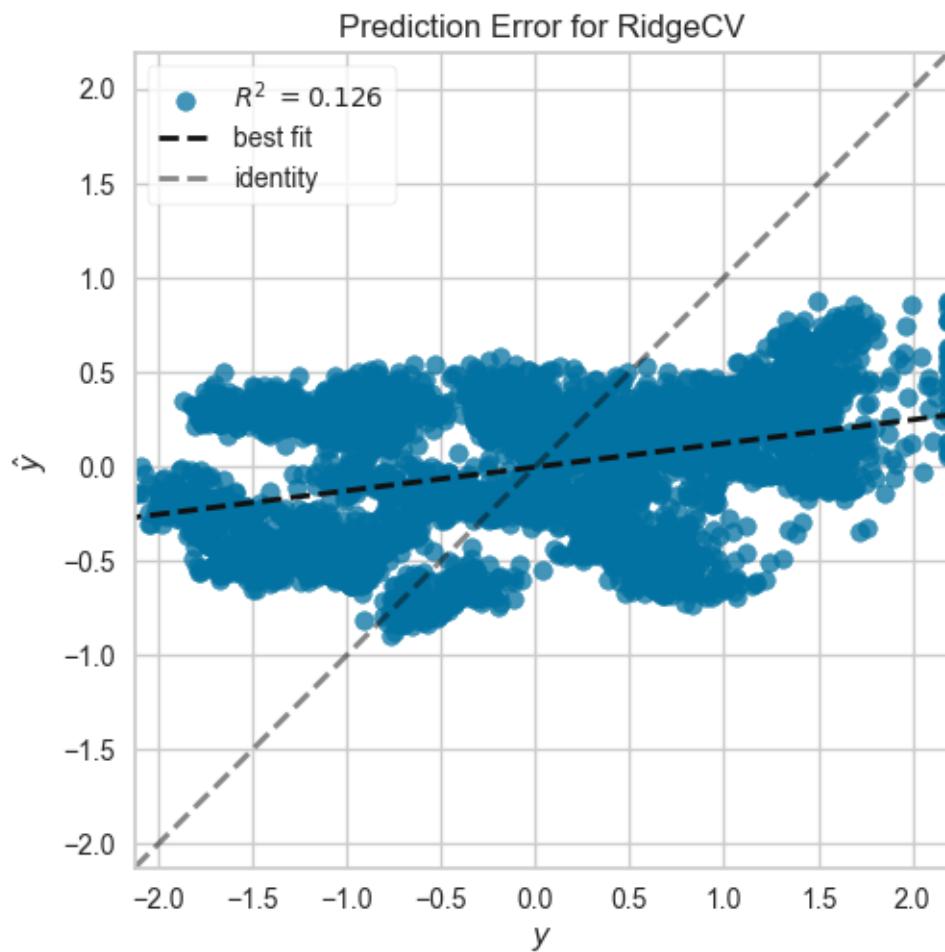


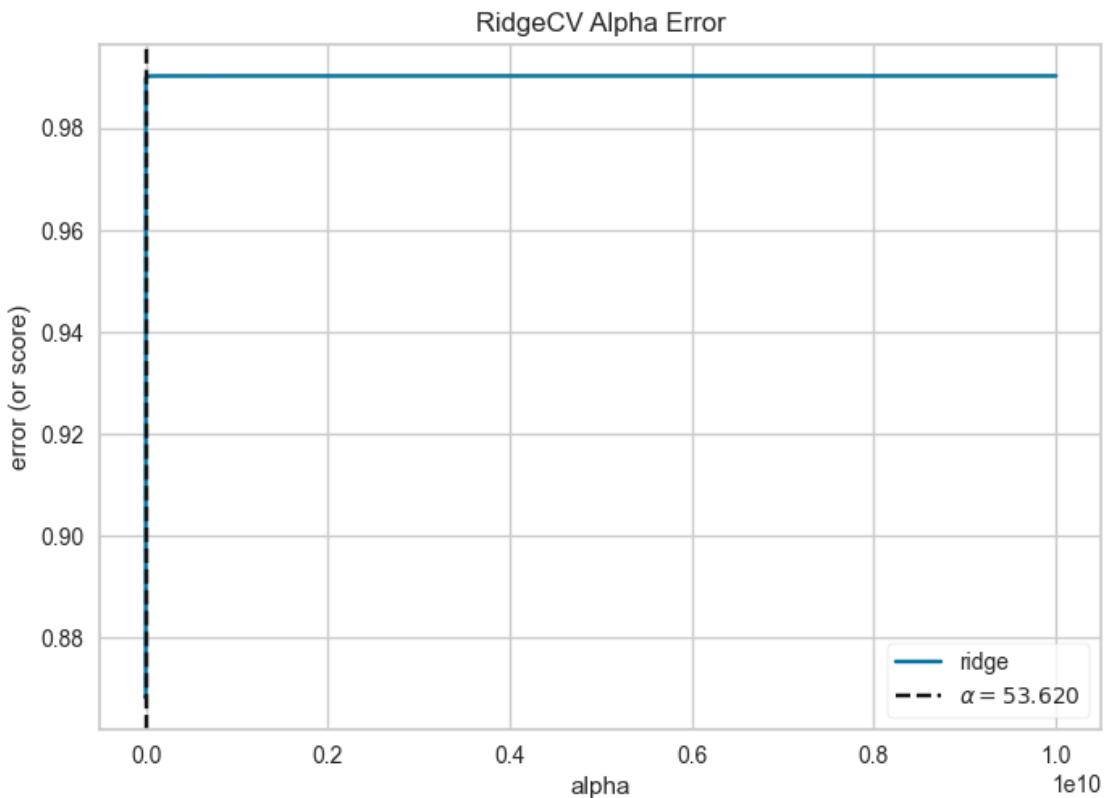


```
[63]: train_and_evaluate_model(RidgeCV())
```

```
Mean Absolute Percentage Error(MAPE): 1.8583366137333945
Root Mean Squared Error(RMSE): 0.9429127460413069
R2 Score: 0.1314683581433027
```

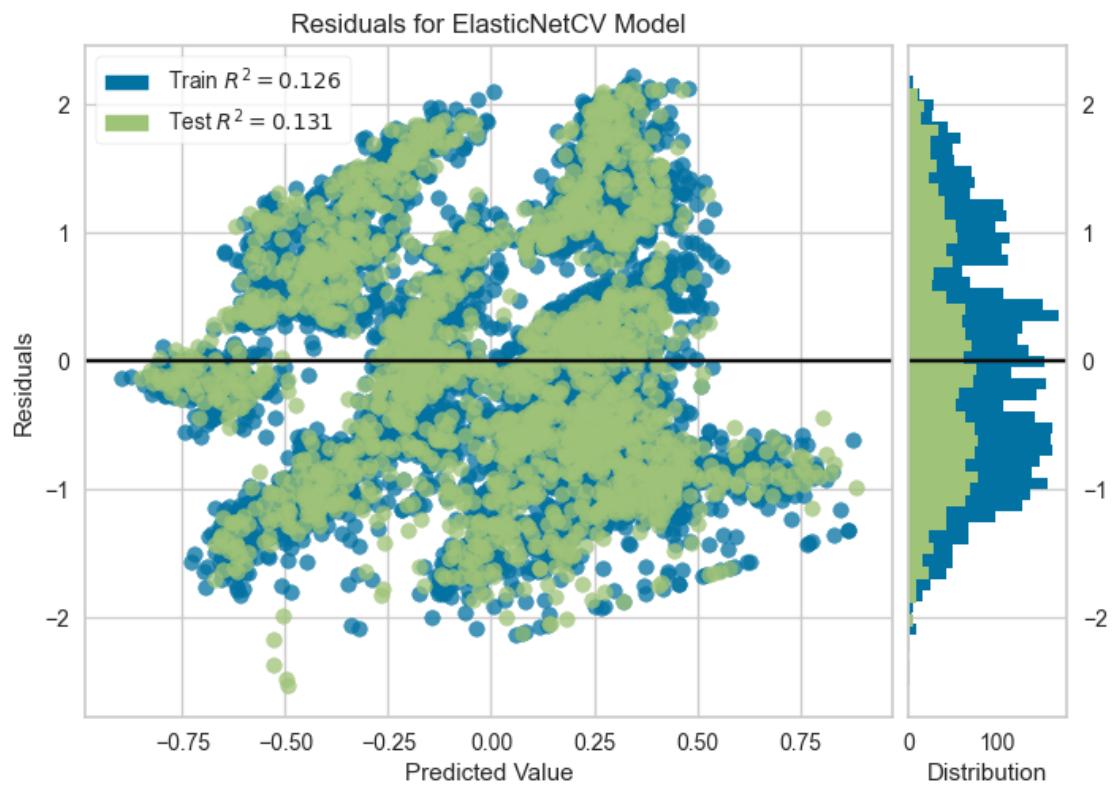


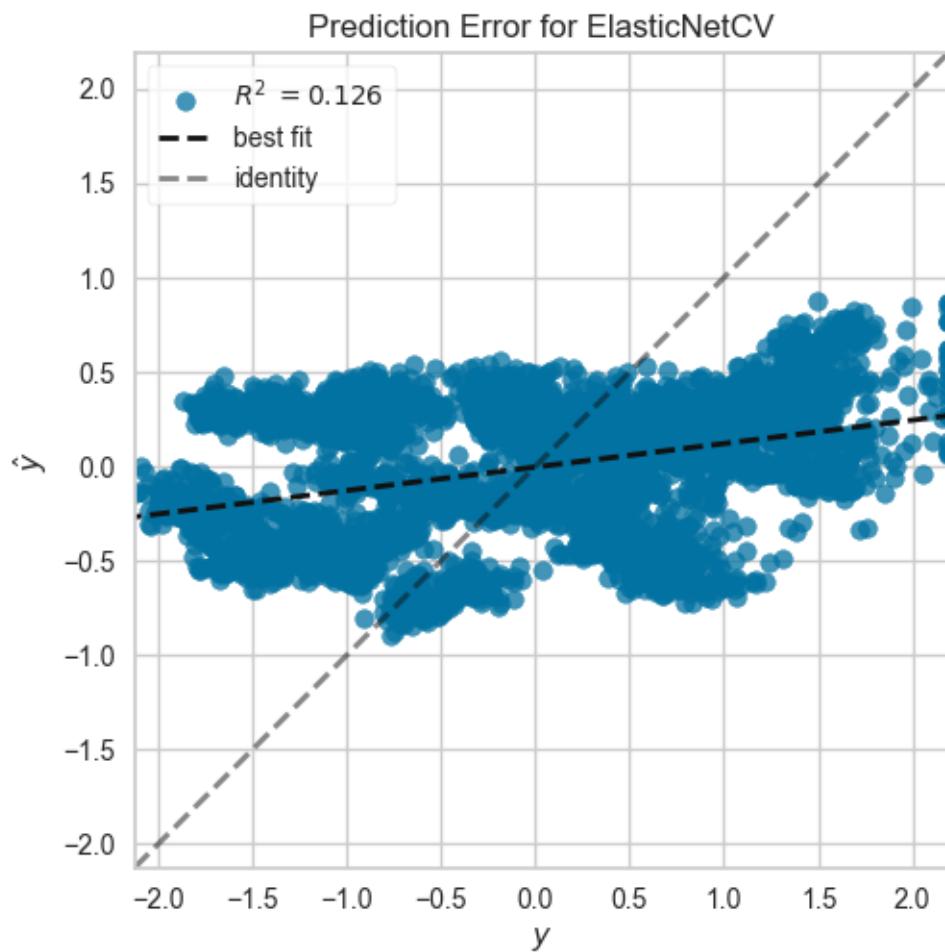


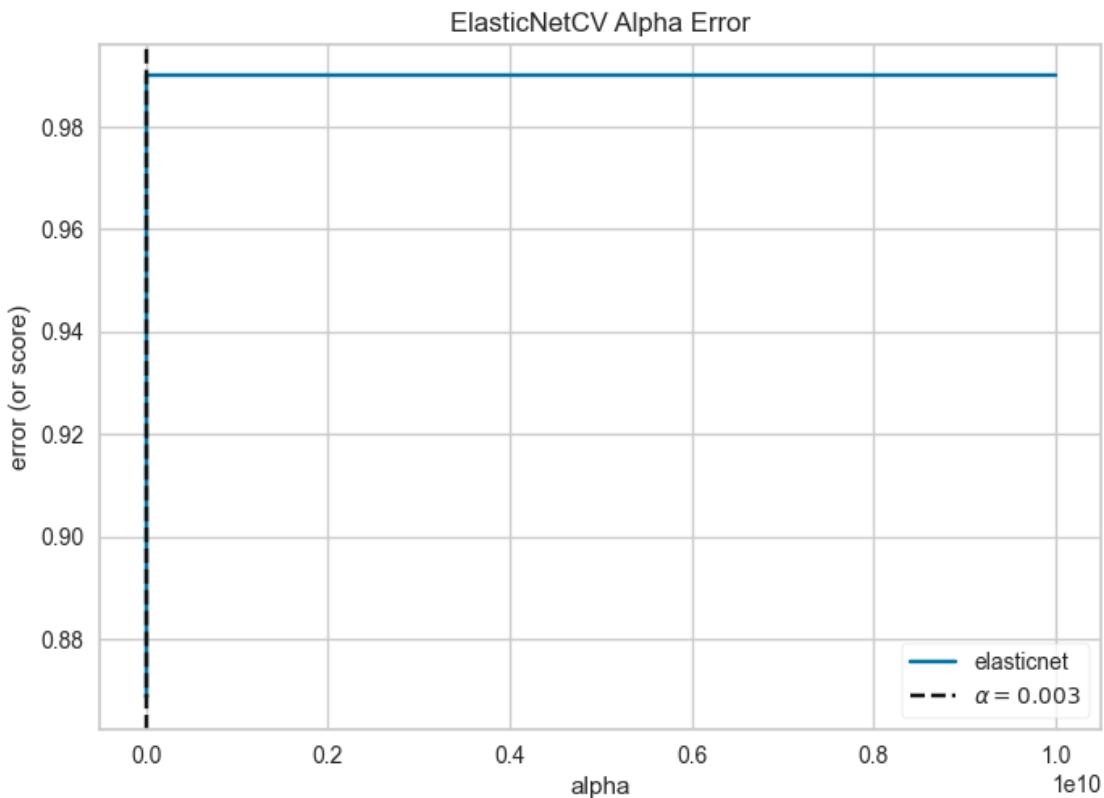


```
[64]: train_and_evaluate_model(ElasticNetCV())
```

```
Mean Absolute Percentage Error(MAPE): 1.8694173949342796
Root Mean Squared Error(RMSE): 0.9431117395338787
R2 Score: 0.13110172743880832
```

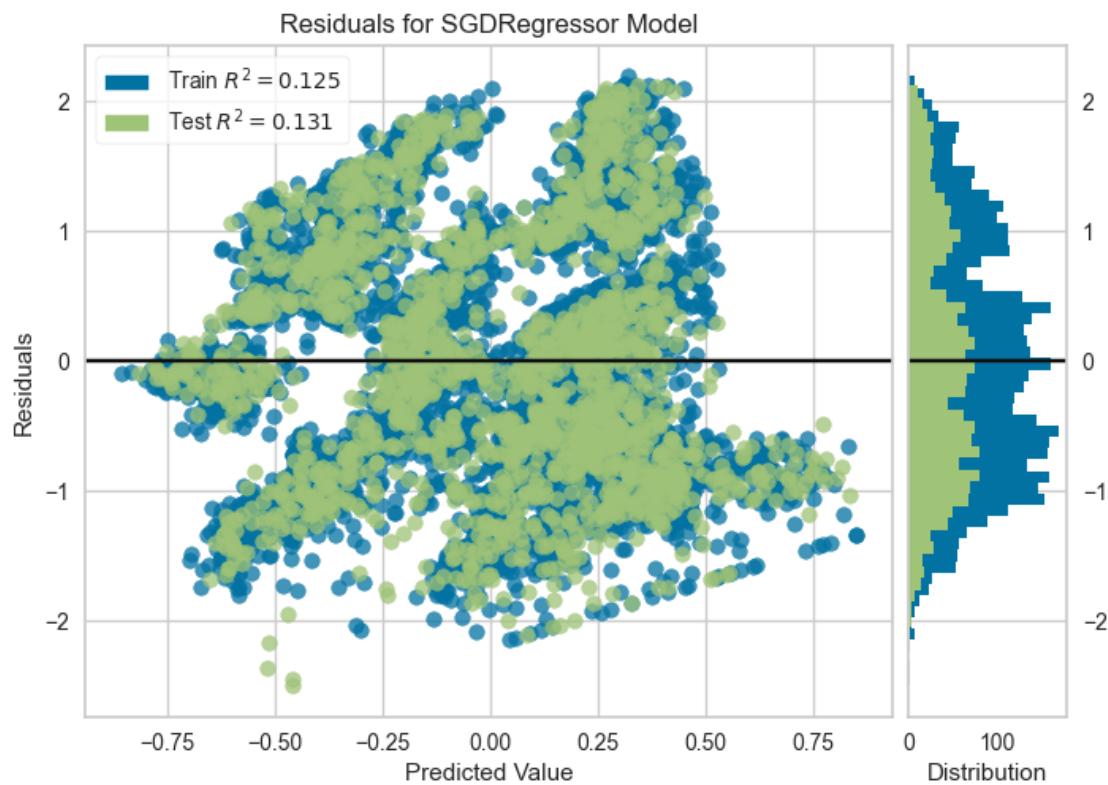


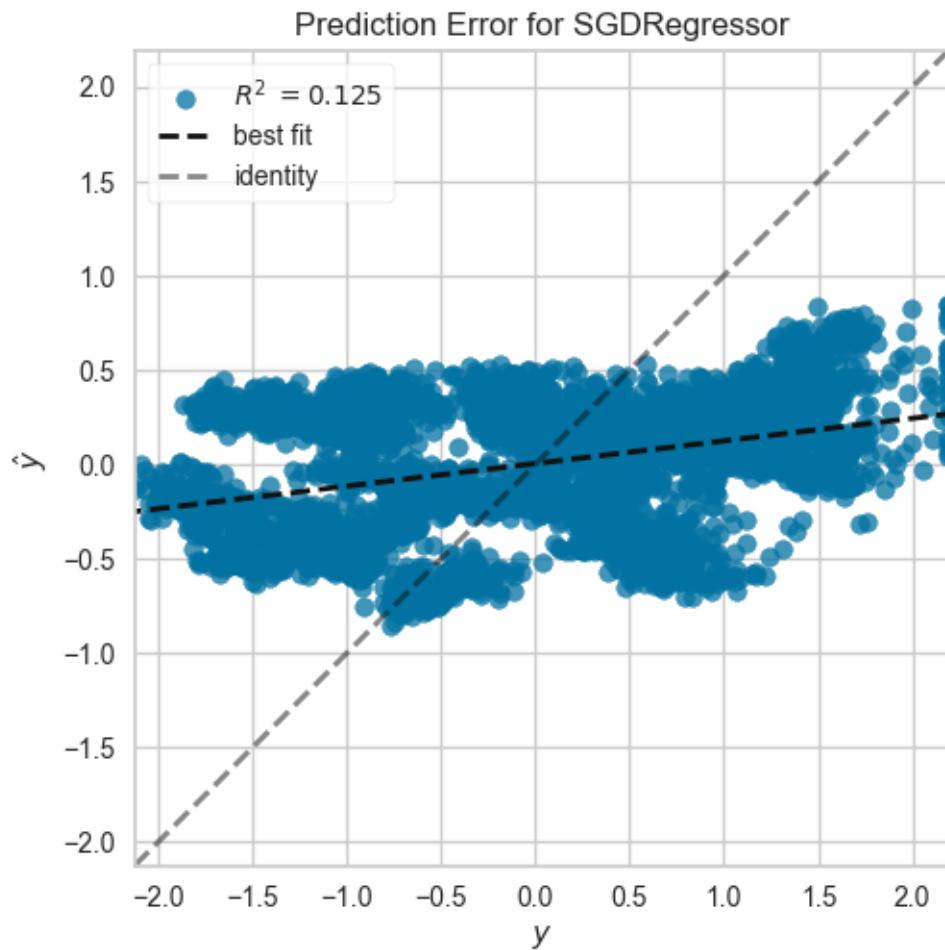




```
[65]: train_and_evaluate_model(SGDRegressor())
```

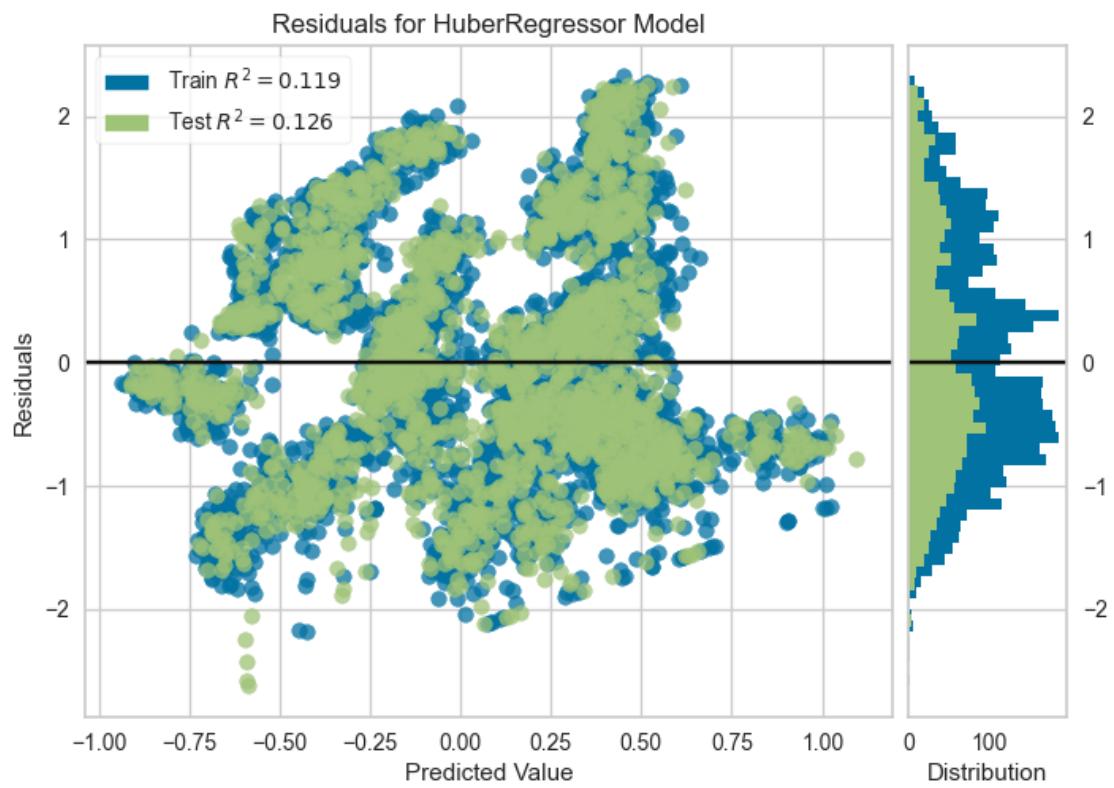
```
Mean Absolute Percentage Error(MAPE): 1.8248079631075012
Root Mean Squared Error(RMSE): 0.9429675988096315
R2 Score: 0.13136730372257965
```

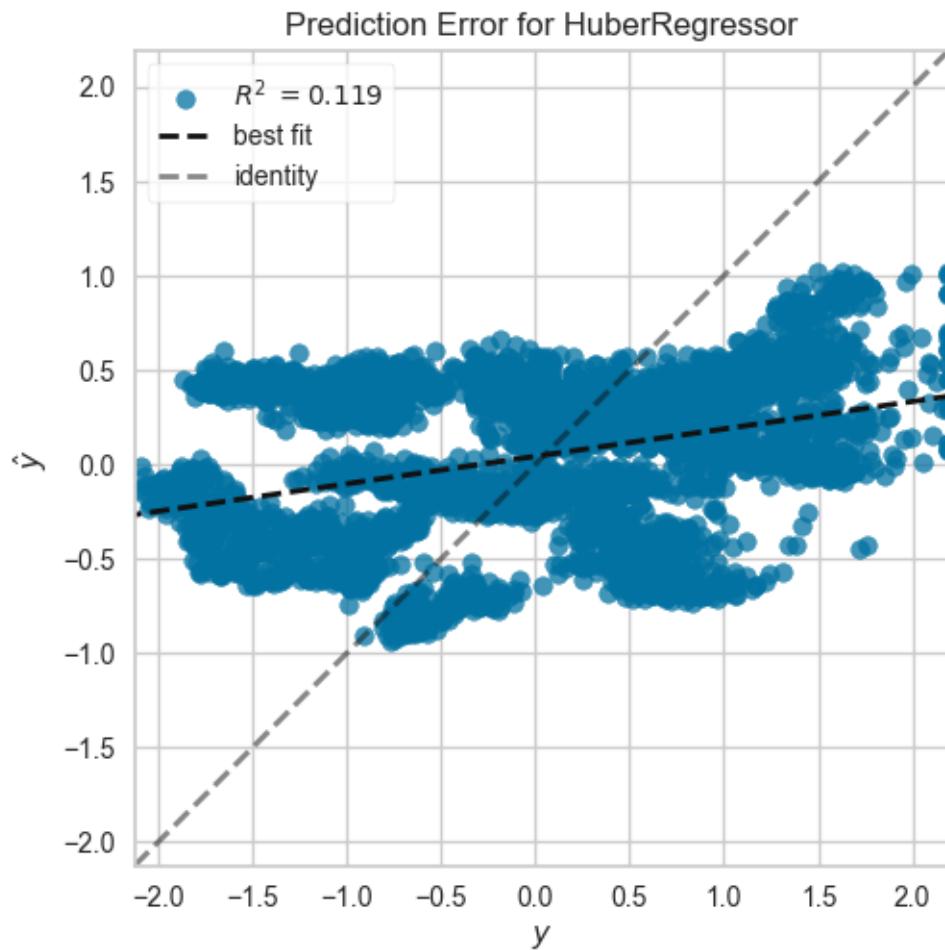




```
[66]: train_and_evaluate_model(HuberRegressor())
```

```
Mean Absolute Percentage Error(MAPE): 2.047510142946776
Root Mean Squared Error(RMSE): 0.9456807950172547
R2 Score: 0.12636148697284888
```



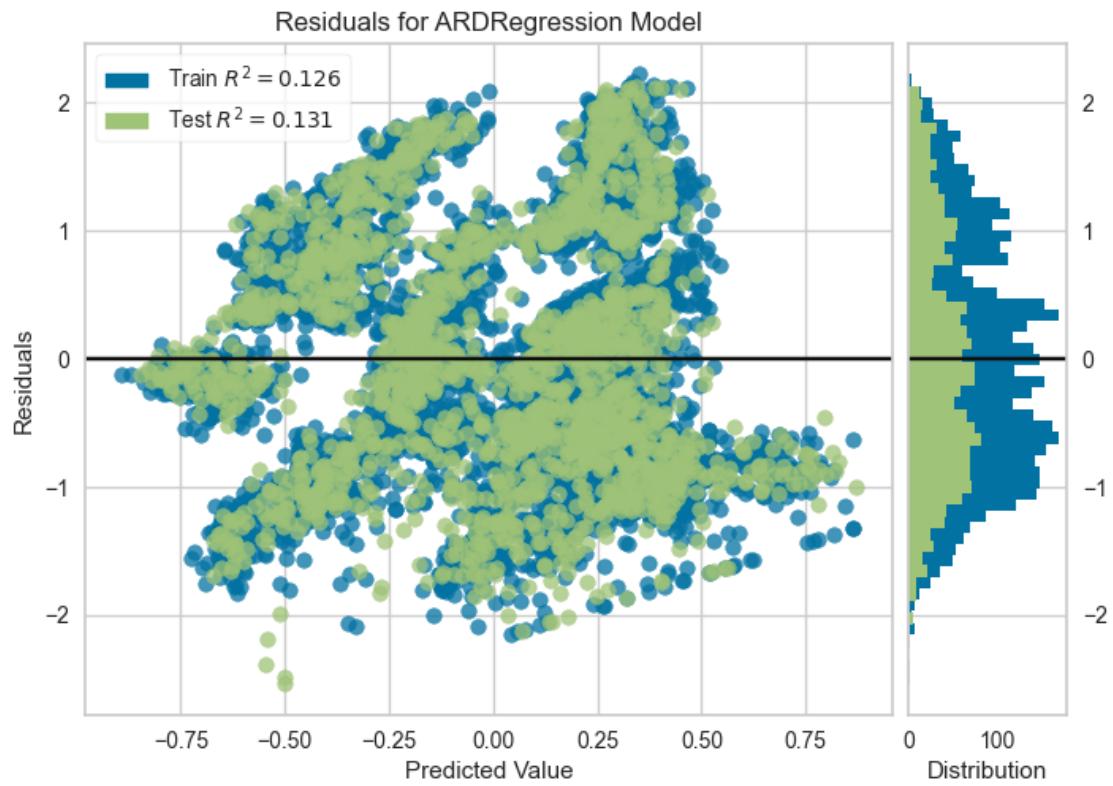


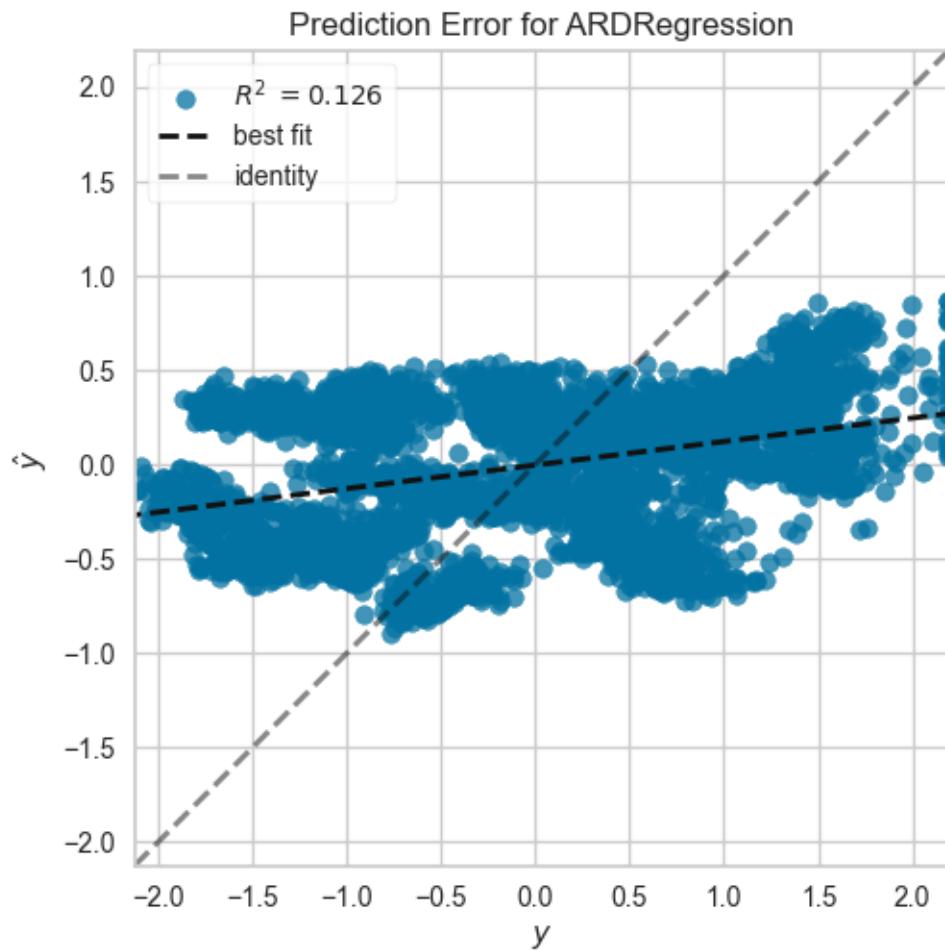
```
[67]: train_and_evaluate_model(ARDRegression())
```

Mean Absolute Percentage Error(MAPE): 1.8962842173507548

Root Mean Squared Error(RMSE): 0.9431246081769947

R2 Score: 0.13107801525797047



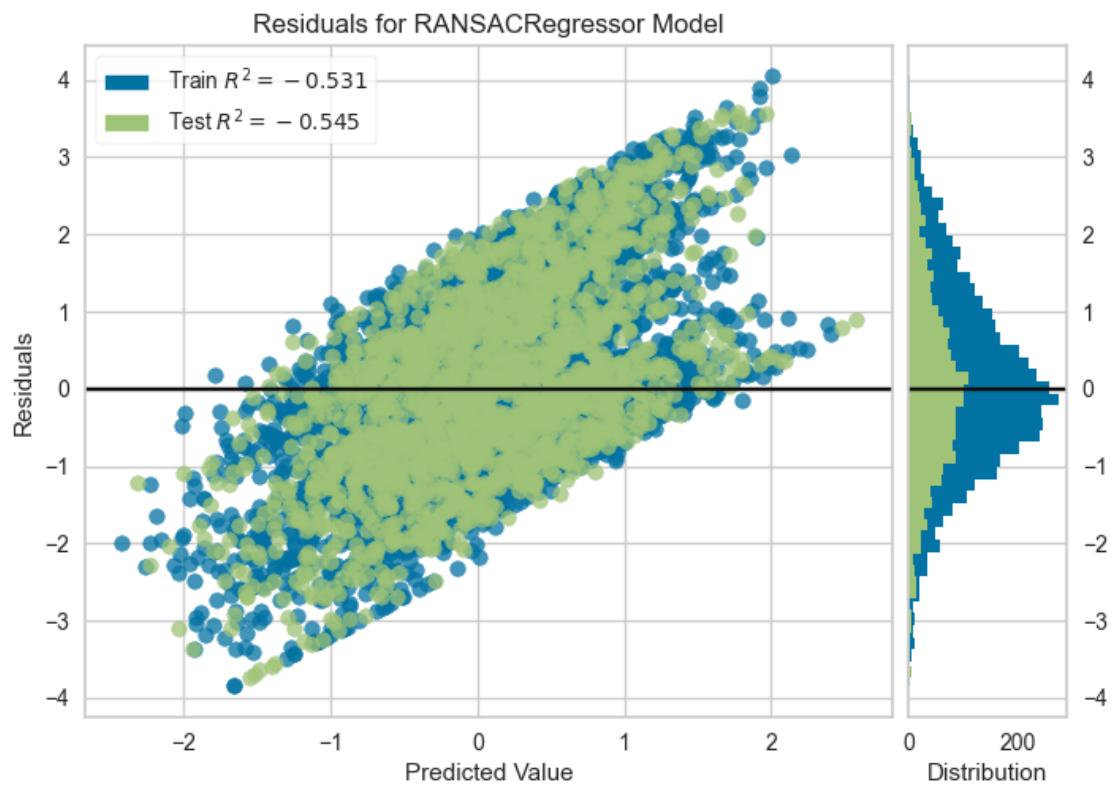


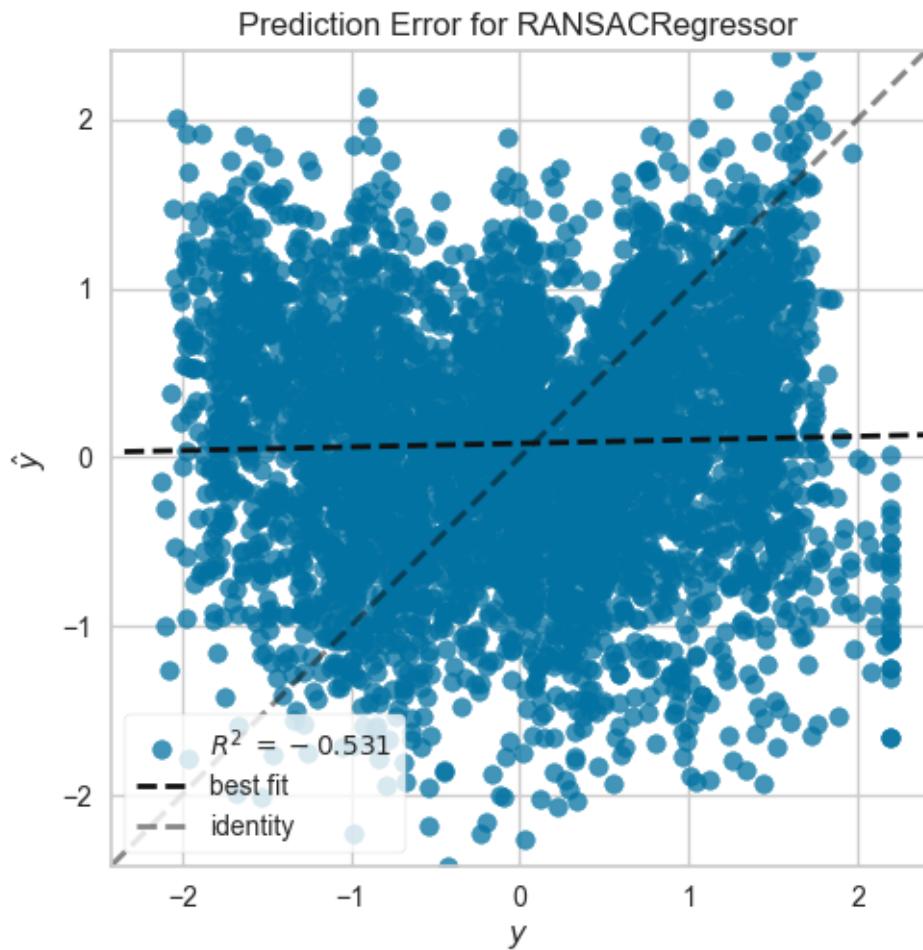
```
[68]: train_and_evaluate_model(RANSACRegressor())
```

Mean Absolute Percentage Error(MAPE): 4.191927385484206

Root Mean Squared Error(RMSE): 1.2575291396130108

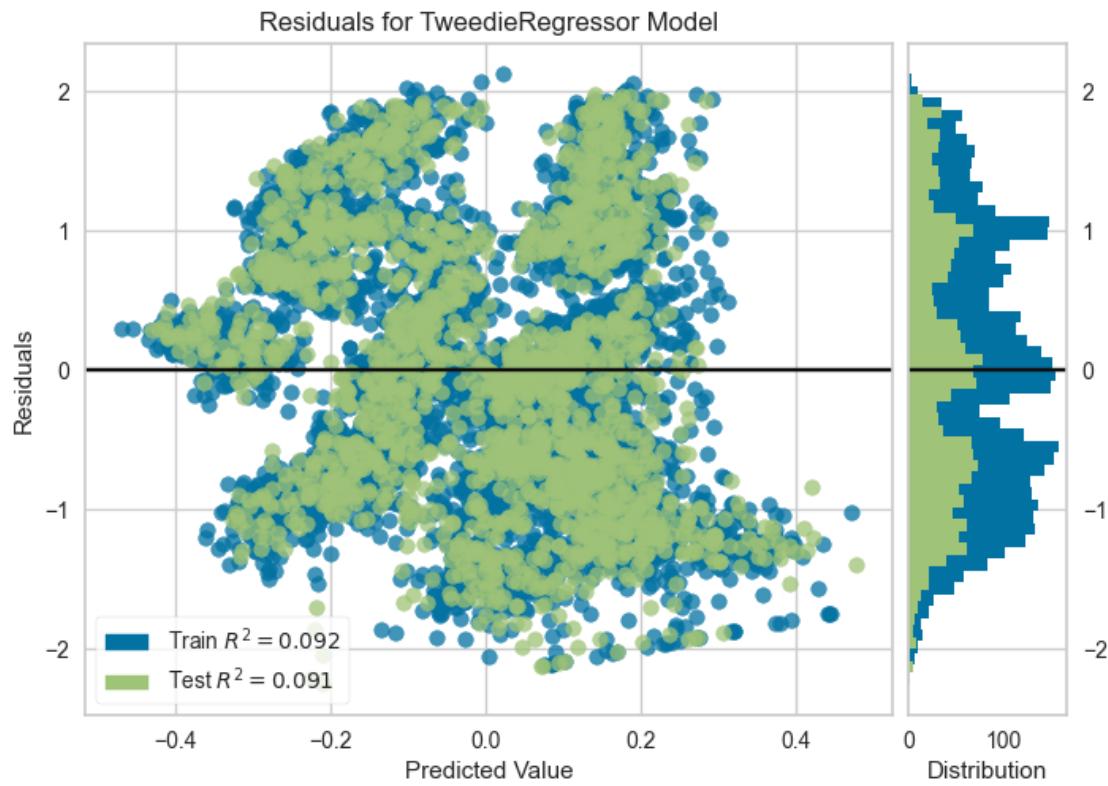
R2 Score: -0.5448230714503632

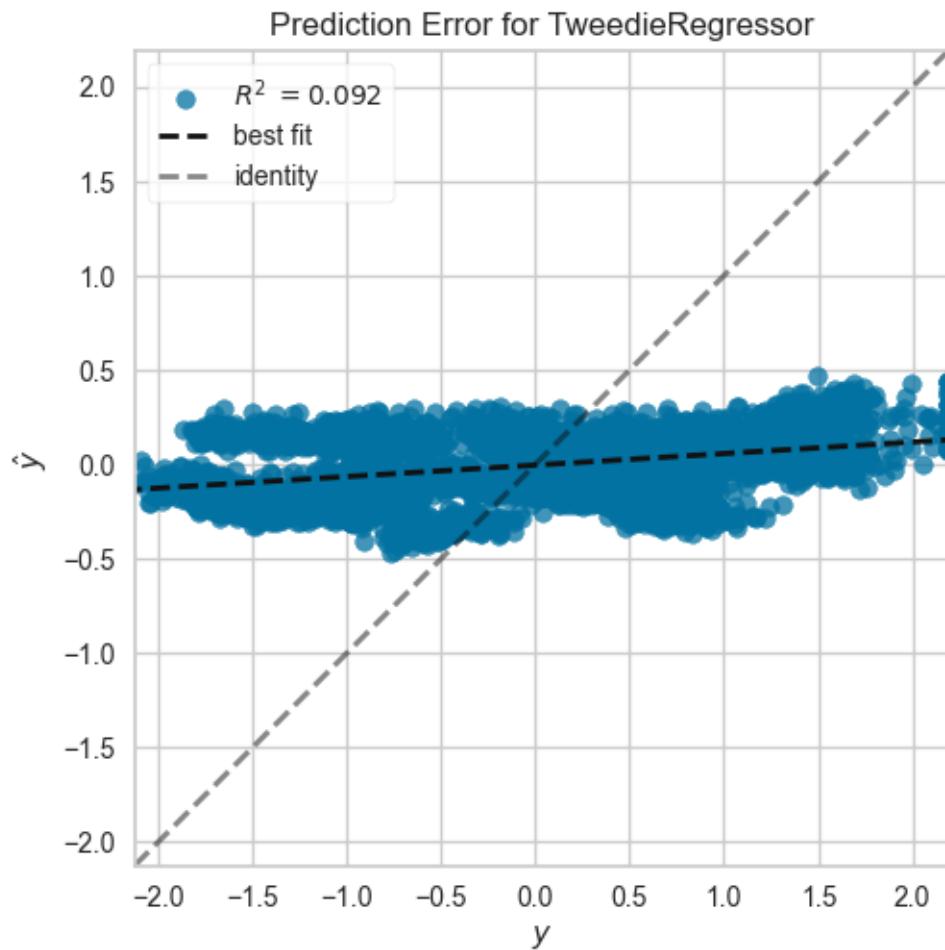




```
[69]: train_and_evaluate_model(TweedieRegressor())
```

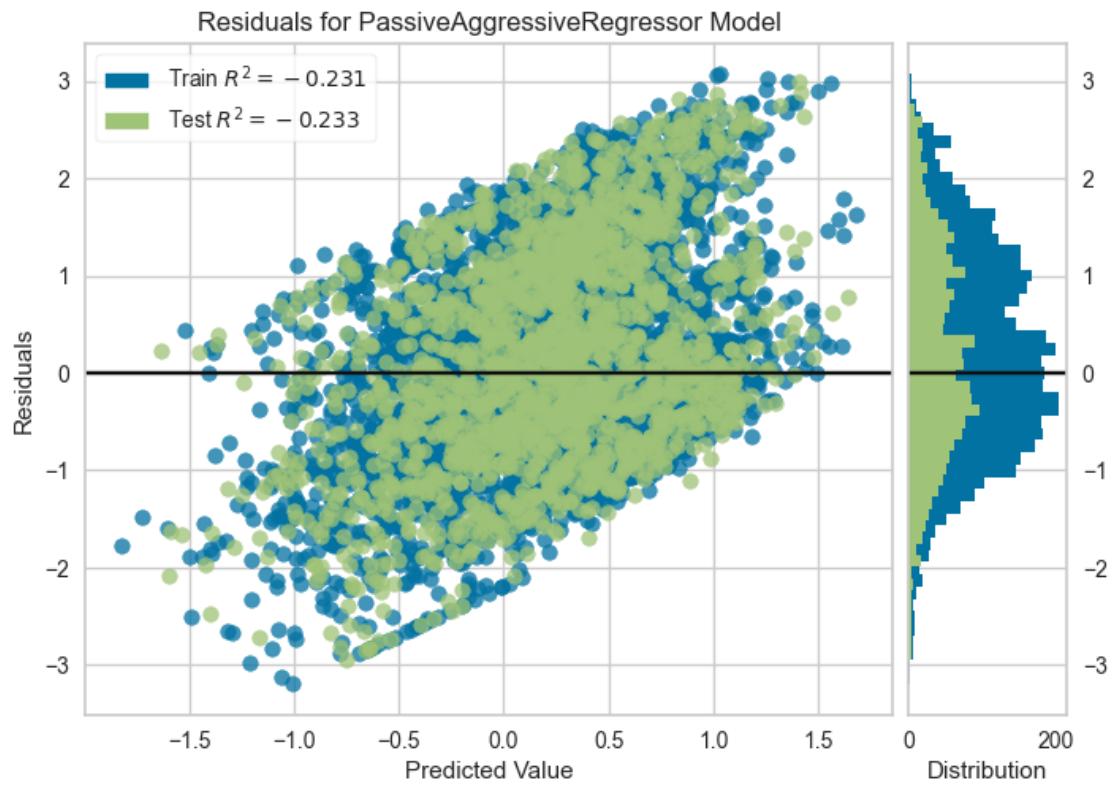
```
Mean Absolute Percentage Error(MAPE): 1.30032699724825
Root Mean Squared Error(RMSE): 0.9645206708385978
R2 Score: 0.09120545570586358
```

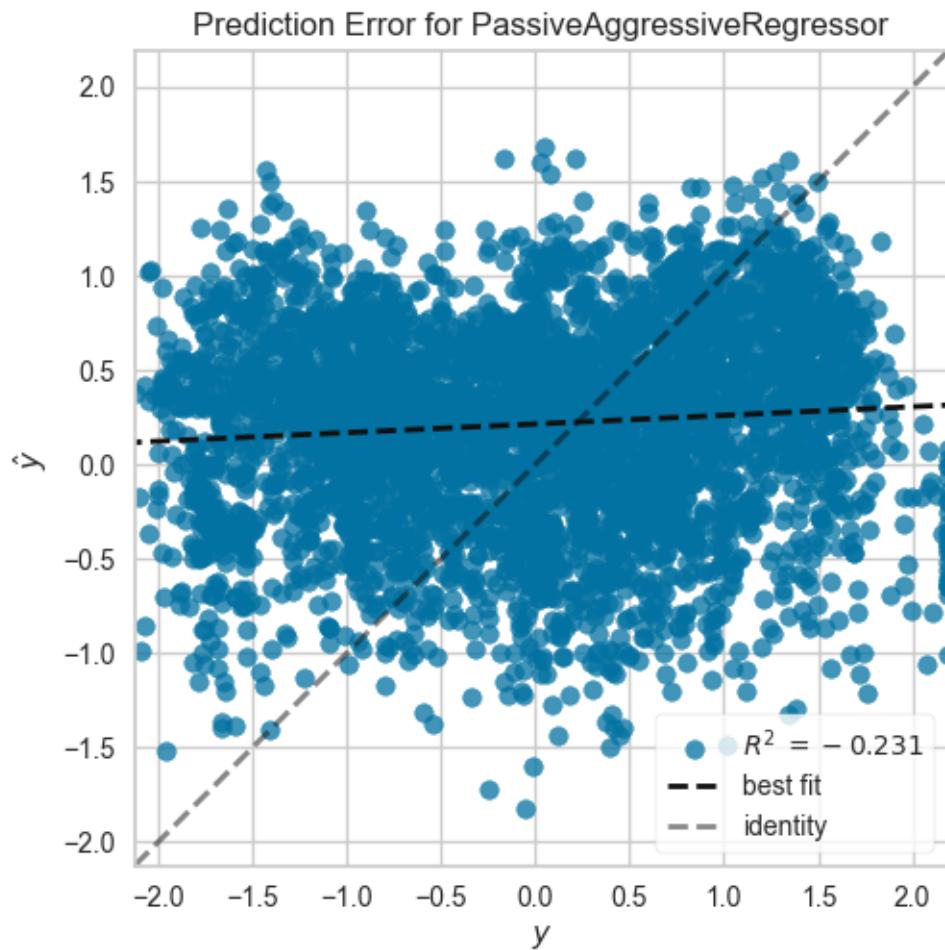




```
[70]: train_and_evaluate_model(PassiveAggressiveRegressor())
```

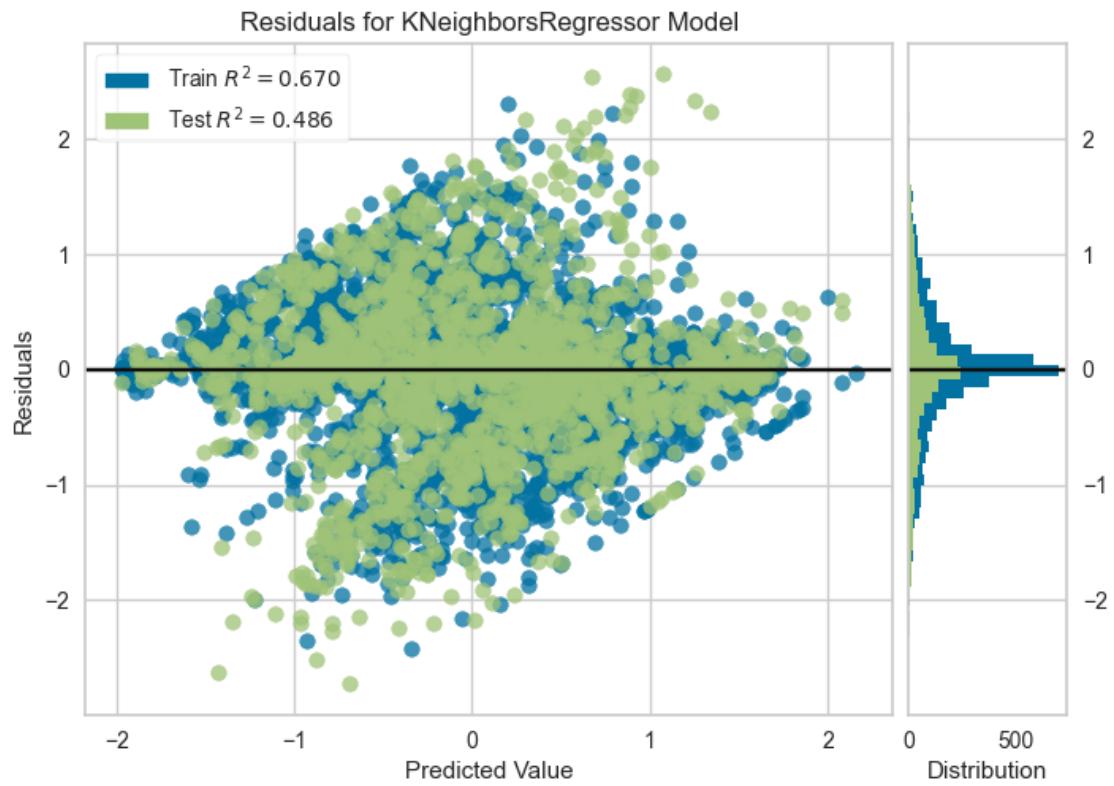
```
Mean Absolute Percentage Error(MAPE): 5.203633425717179
Root Mean Squared Error(RMSE): 1.1235222151525137
R2 Score: -0.23312175640425492
```

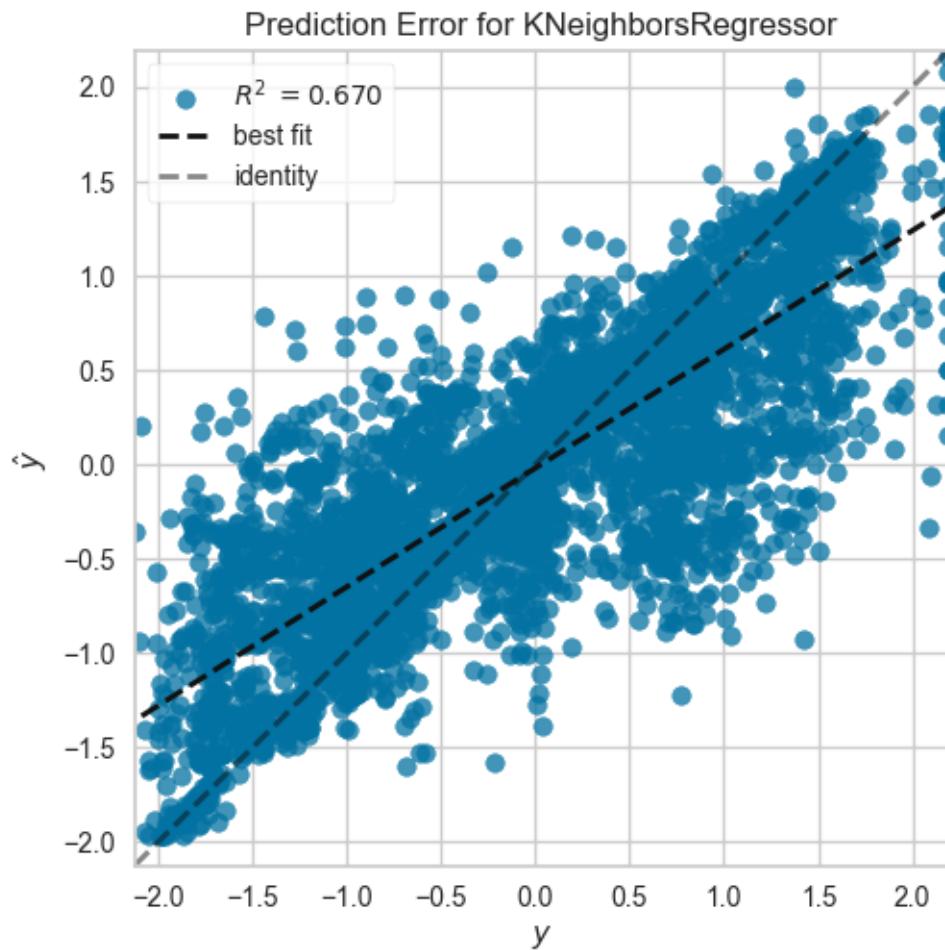




```
[71]: train_and_evaluate_model(KNeighborsRegressor())
```

```
Mean Absolute Percentage Error(MAPE): 1.8324259921256374
Root Mean Squared Error(RMSE): 0.725657170052076
R2 Score: 0.485594487792688
```



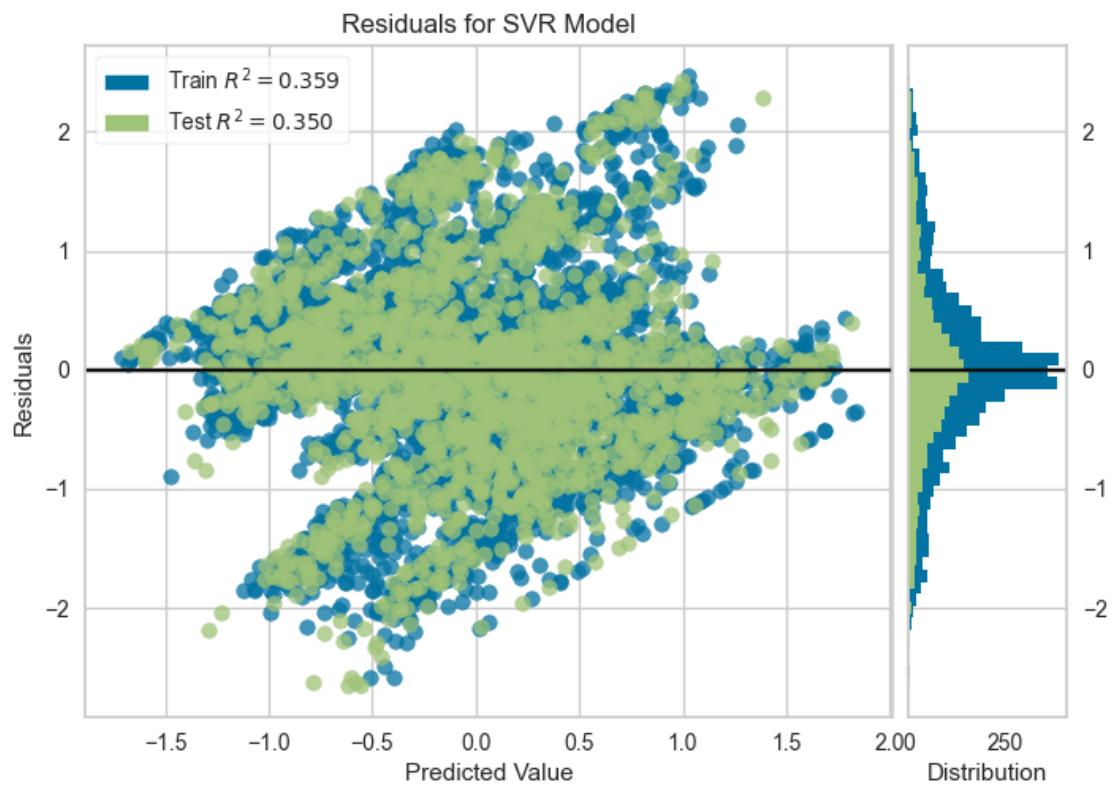


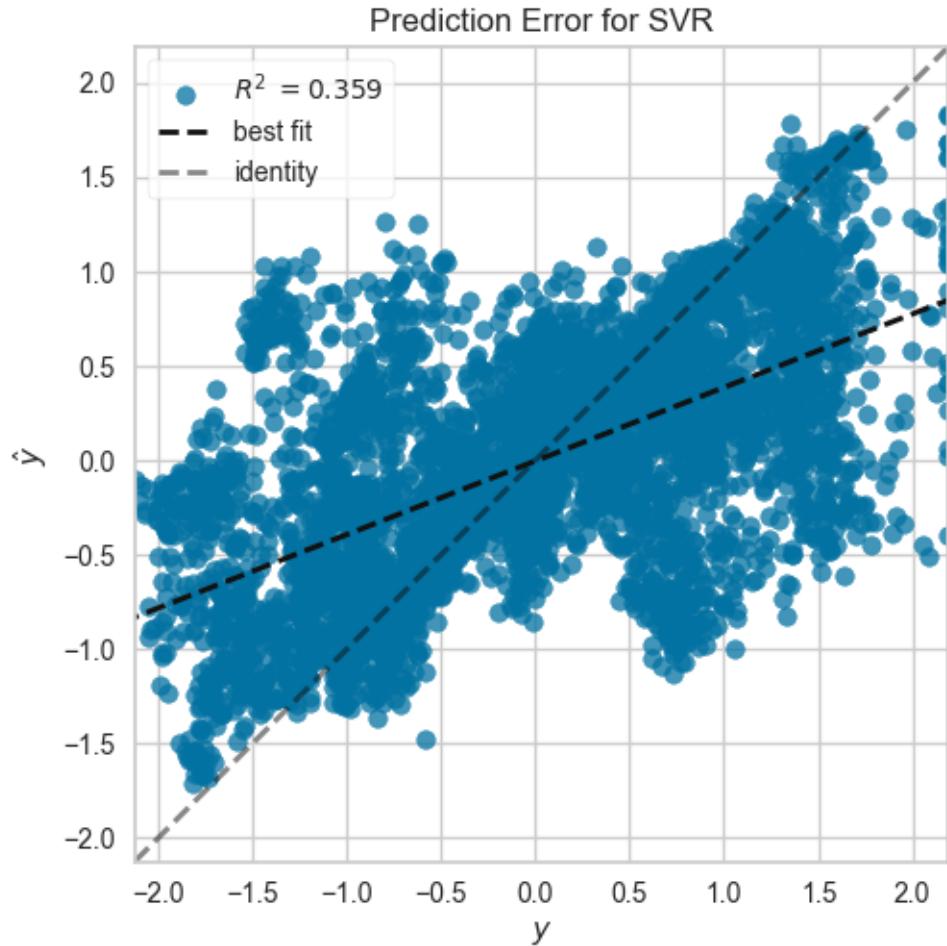
```
[72]: train_and_evaluate_model(SVR())
```

Mean Absolute Percentage Error(MAPE): 1.7786440447853233

Root Mean Squared Error(RMSE): 0.8159376019038274

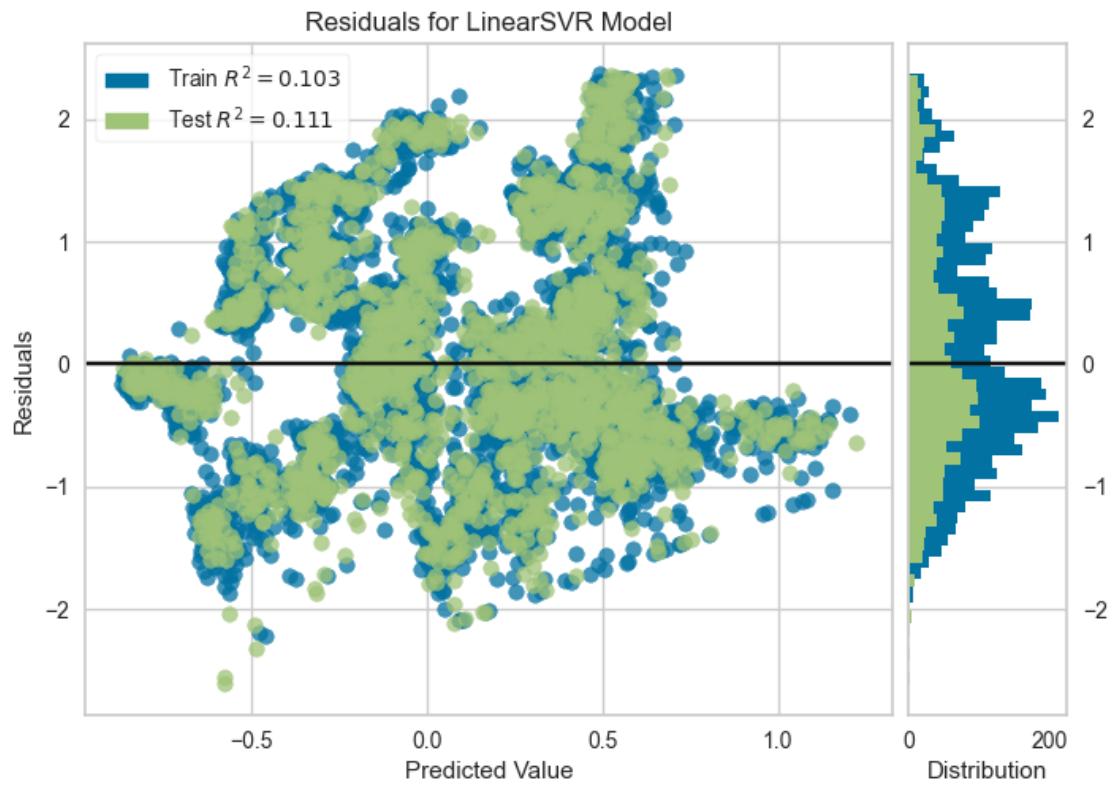
R2 Score: 0.3496359488709865

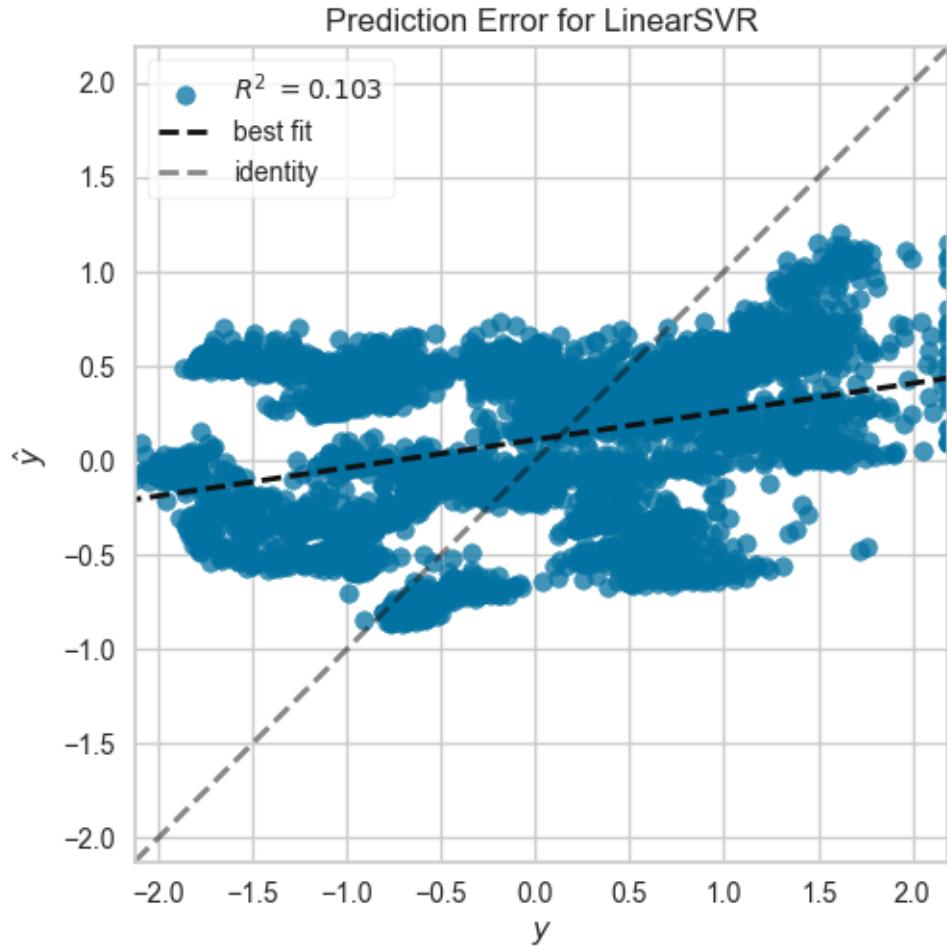




```
[73]: train_and_evaluate_model(LinearSVR())
```

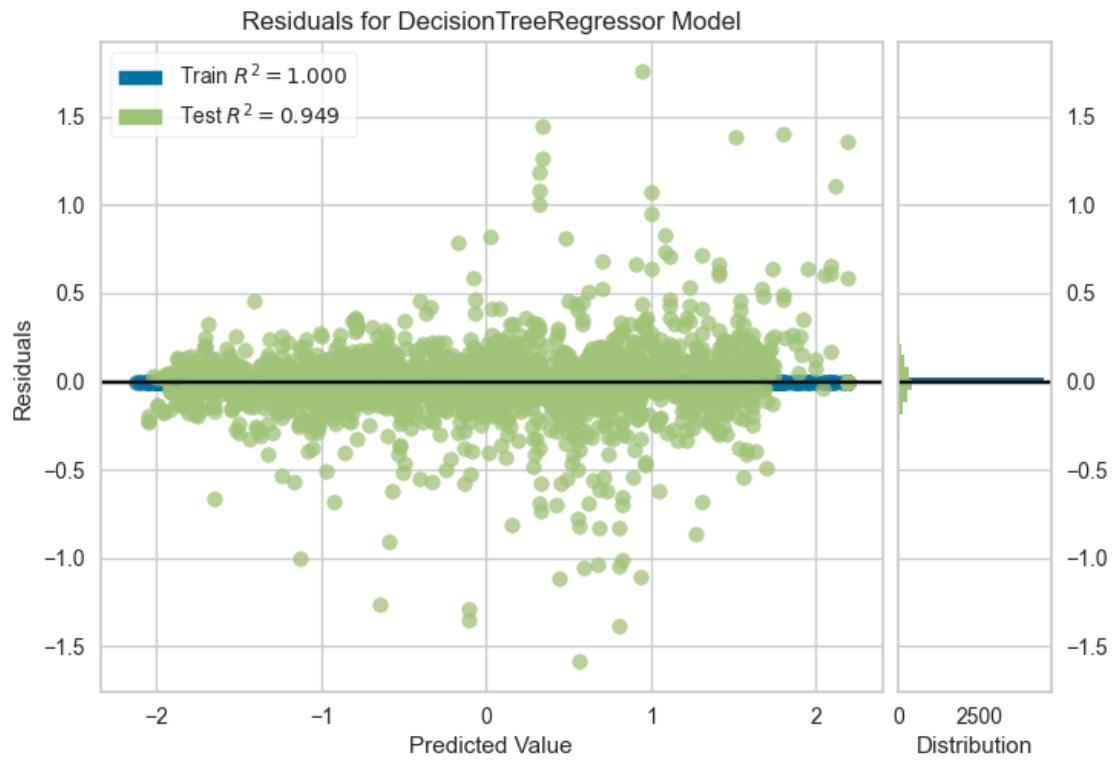
```
Mean Absolute Percentage Error(MAPE): 2.058692310166648
Root Mean Squared Error(RMSE): 0.9537043048035928
R2 Score: 0.11147404624017132
```

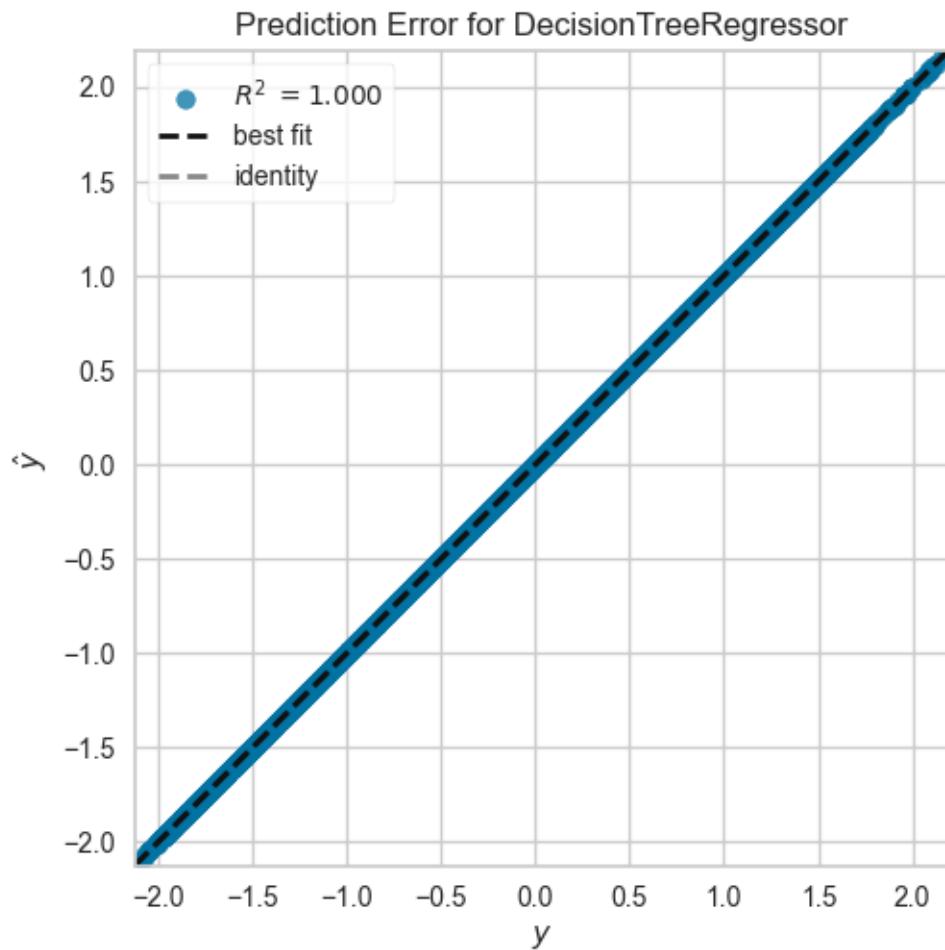




```
[74]: train_and_evaluate_model(DecisionTreeRegressor())
```

```
Mean Absolute Percentage Error(MAPE): 0.6886173929889536
Root Mean Squared Error(RMSE): 0.22928363764923226
R2 Score: 0.9486442875189682
```



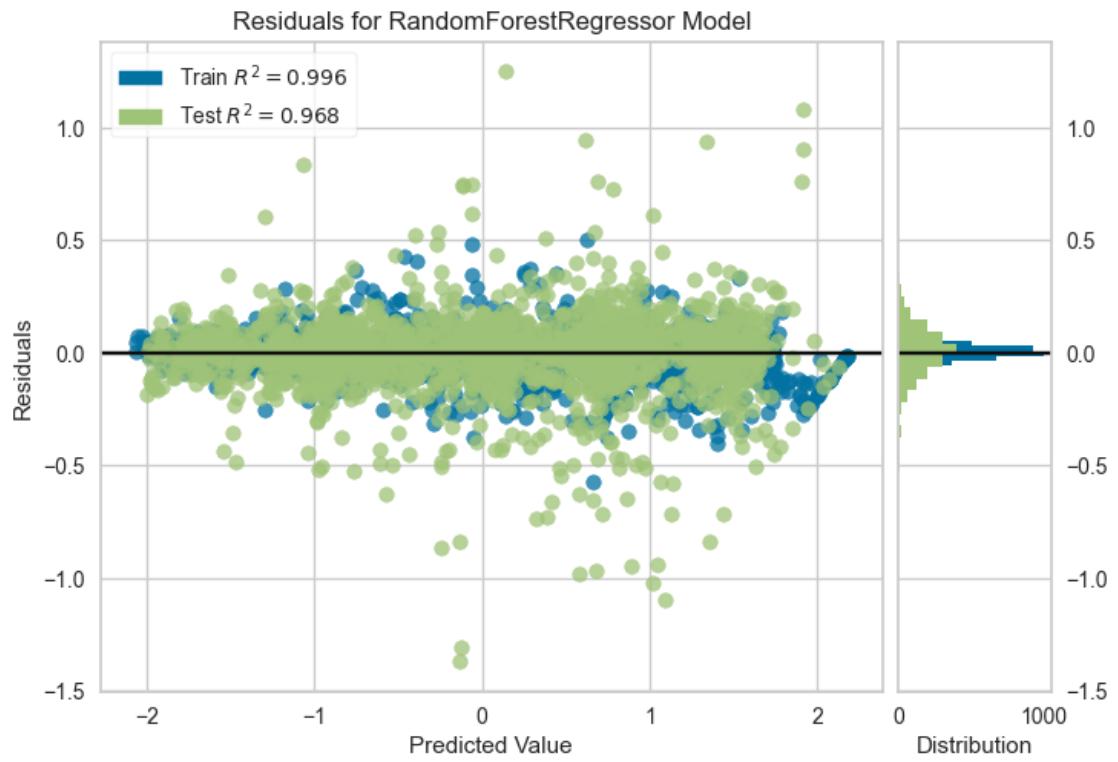


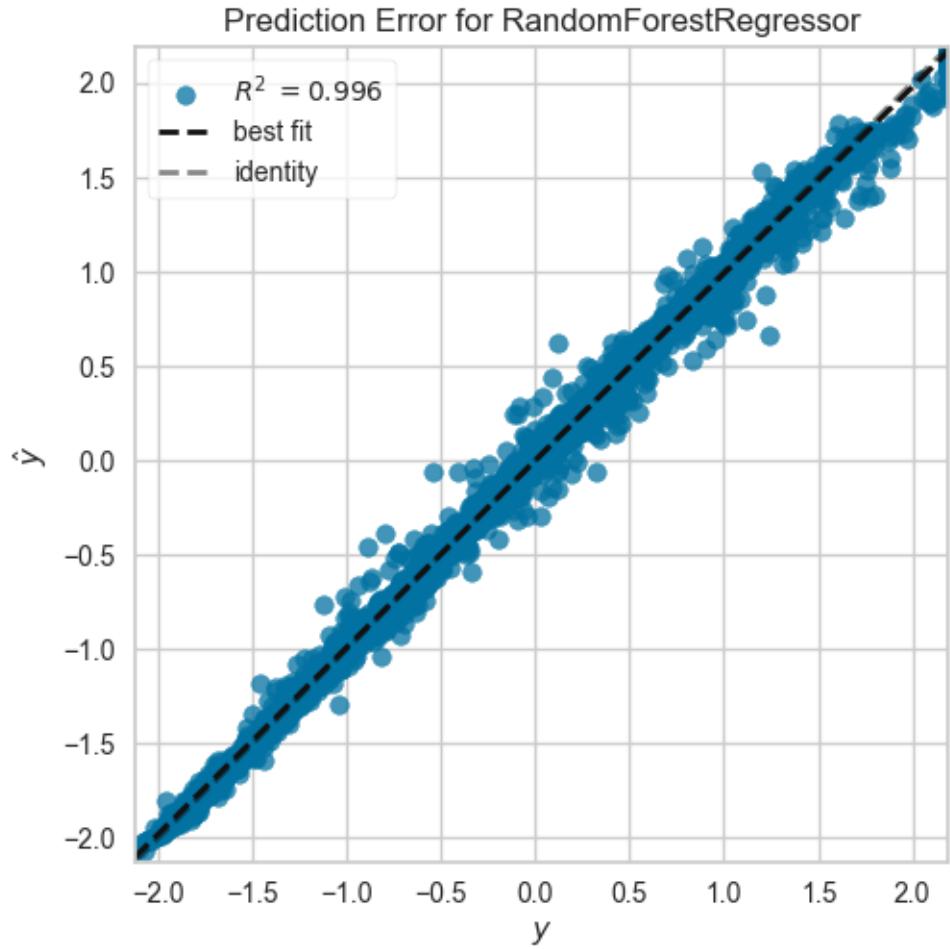
```
[75]: train_and_evaluate_model(RandomForestRegressor())
```

Mean Absolute Percentage Error(MAPE): 0.6316357845454231

Root Mean Squared Error(RMSE): 0.18088904075470633

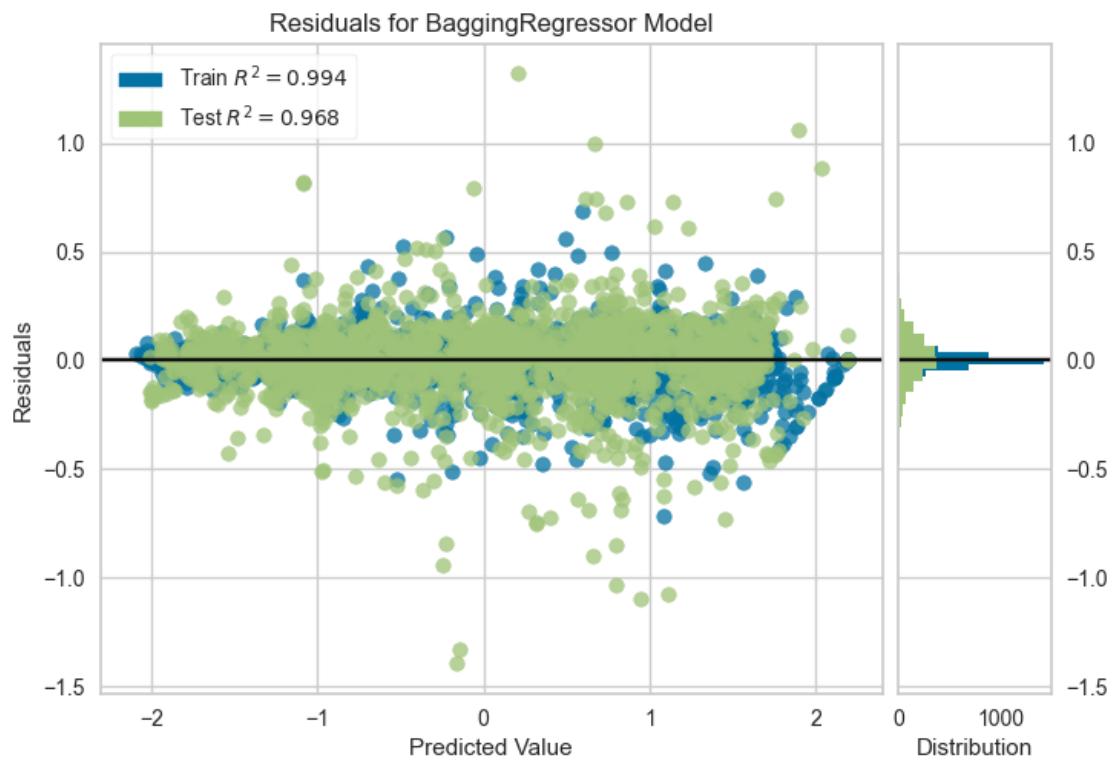
R2 Score: 0.9680355568084091

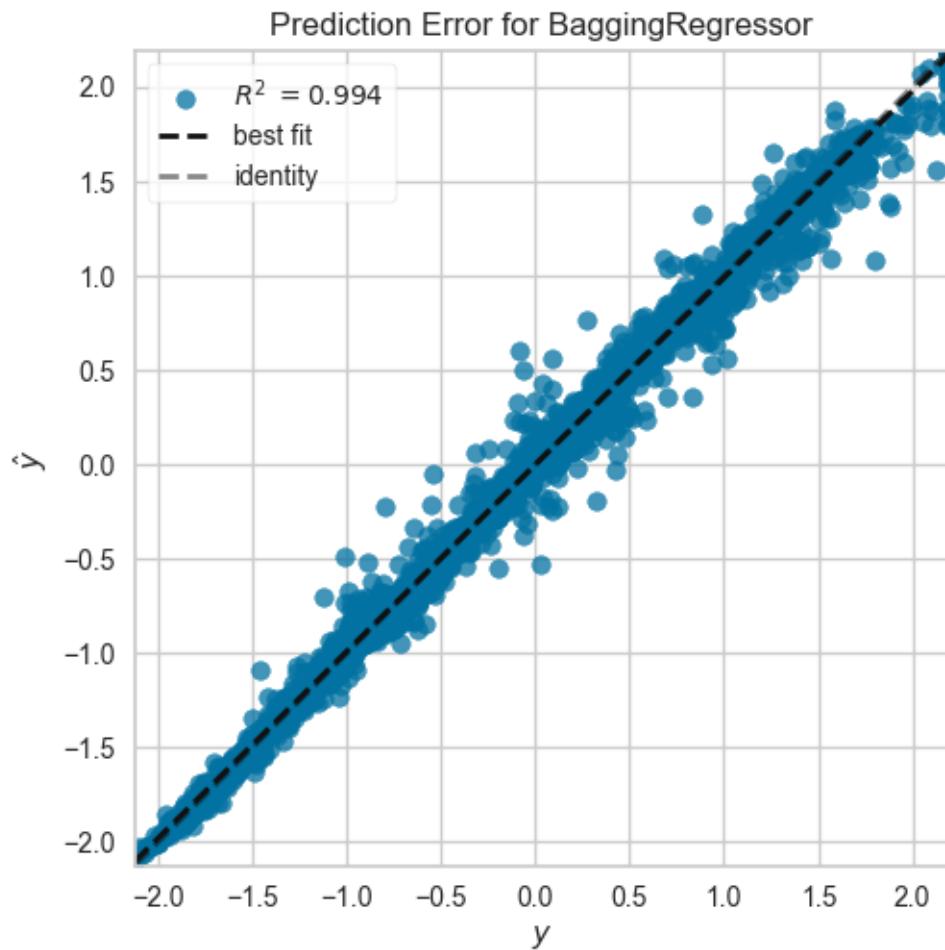




```
[76]: train_and_evaluate_model(BaggingRegressor())
```

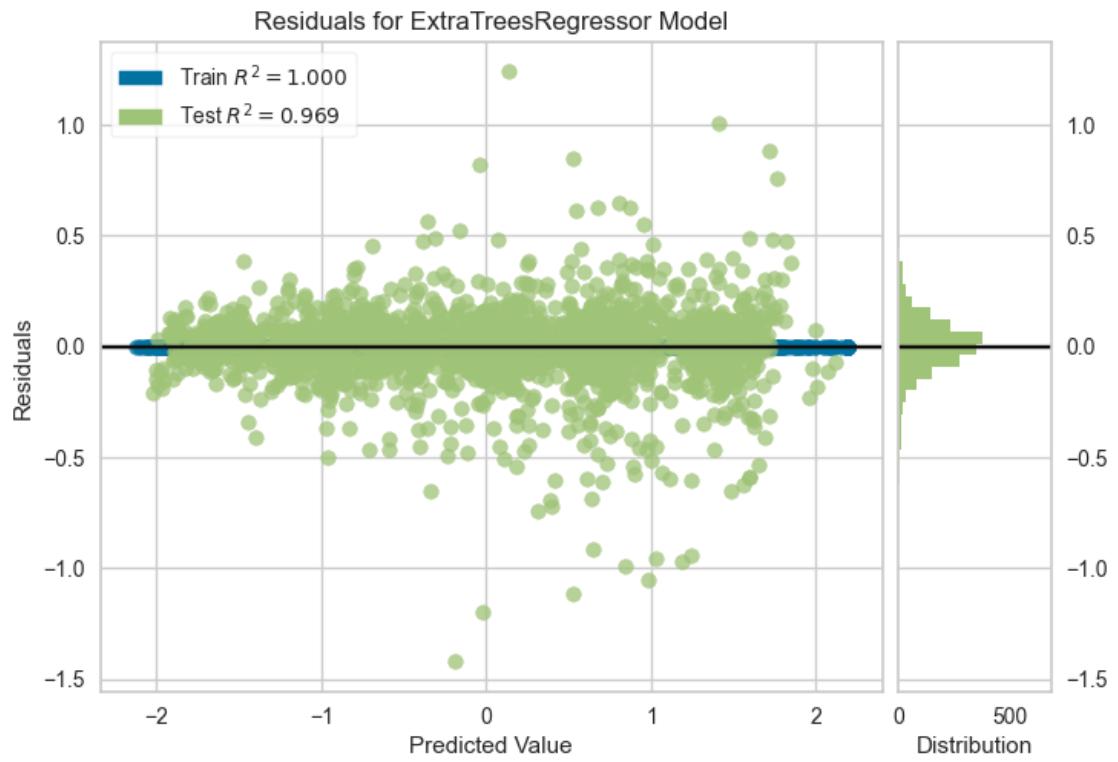
```
Mean Absolute Percentage Error(MAPE): 0.6288842943145447  
Root Mean Squared Error(RMSE): 0.18181709881010605  
R2 Score: 0.9677067258609293
```

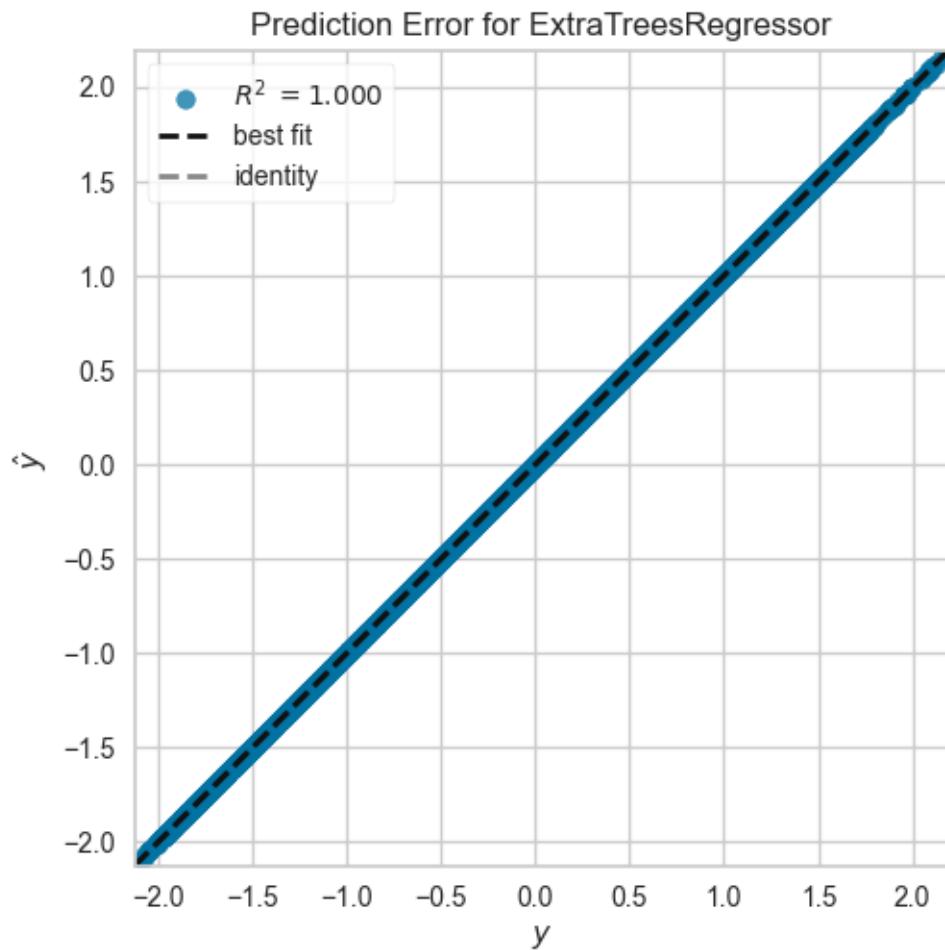




```
[77]: train_and_evaluate_model(ExtraTreesRegressor())
```

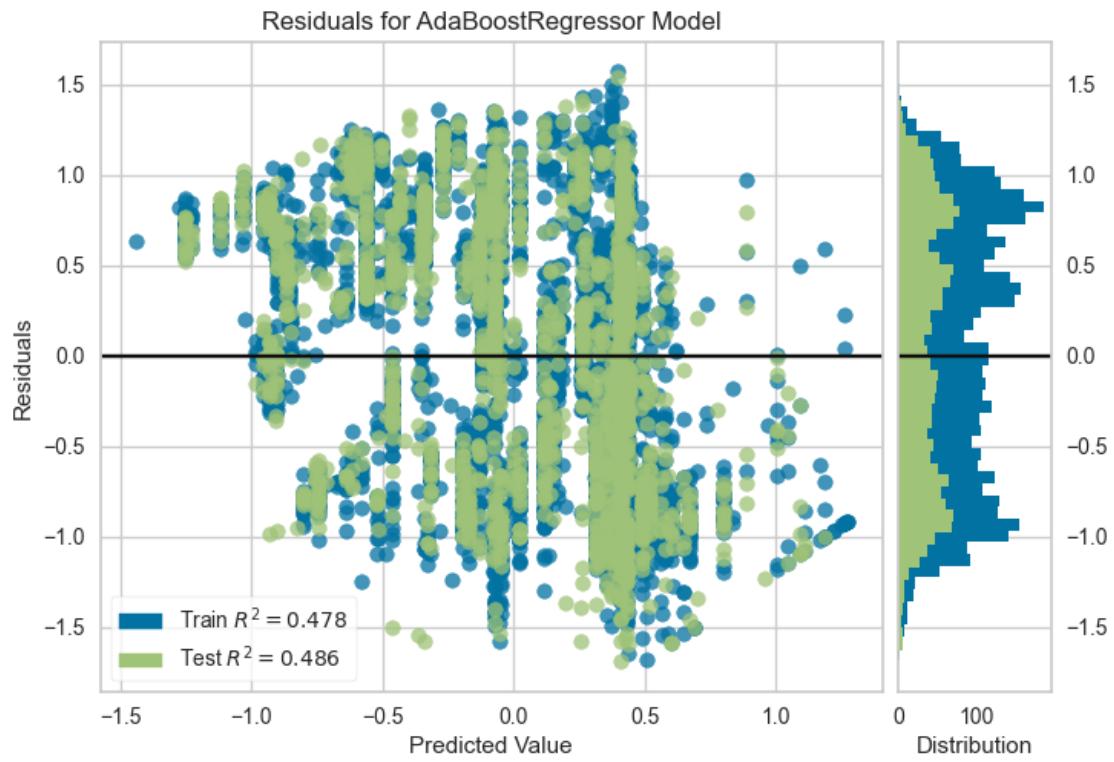
```
Mean Absolute Percentage Error(MAPE): 0.5968460838198617
Root Mean Squared Error(RMSE): 0.17849433508980103
R2 Score: 0.9688762794770702
```

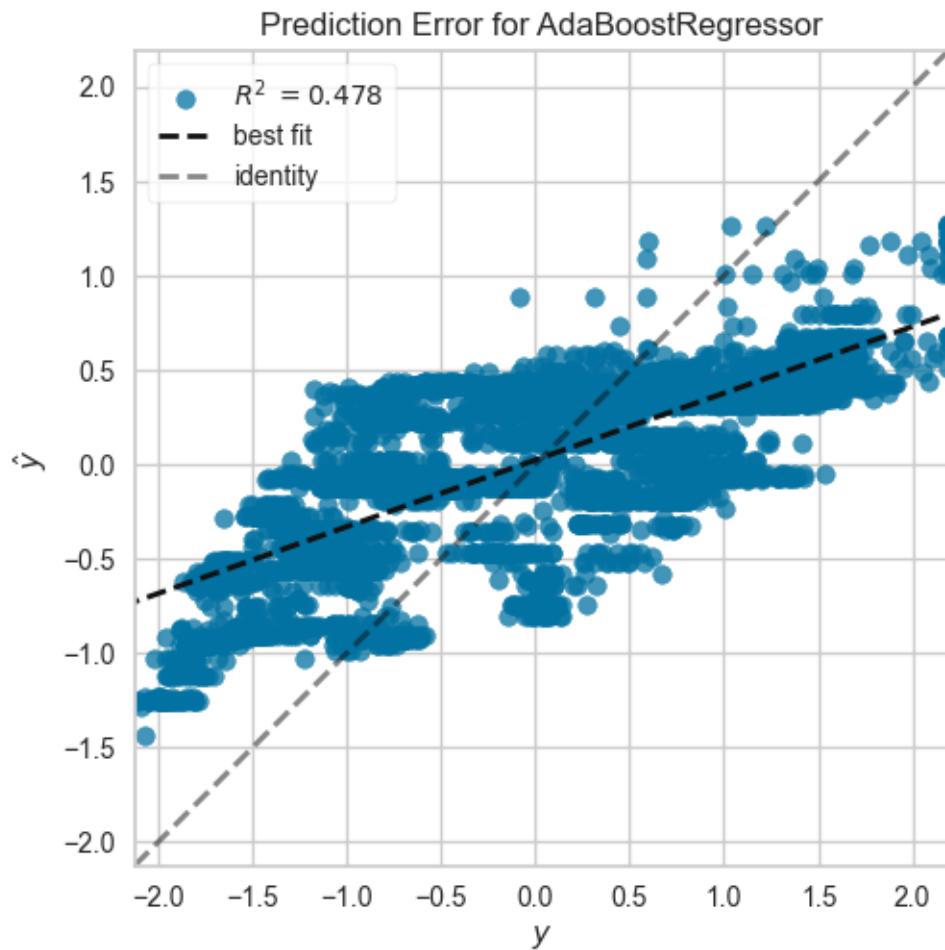




```
[78]: train_and_evaluate_model(AdaBoostRegressor())
```

```
Mean Absolute Percentage Error(MAPE): 2.57532882324867
Root Mean Squared Error(RMSE): 0.7250431693889414
R2 Score: 0.4864646278959307
```

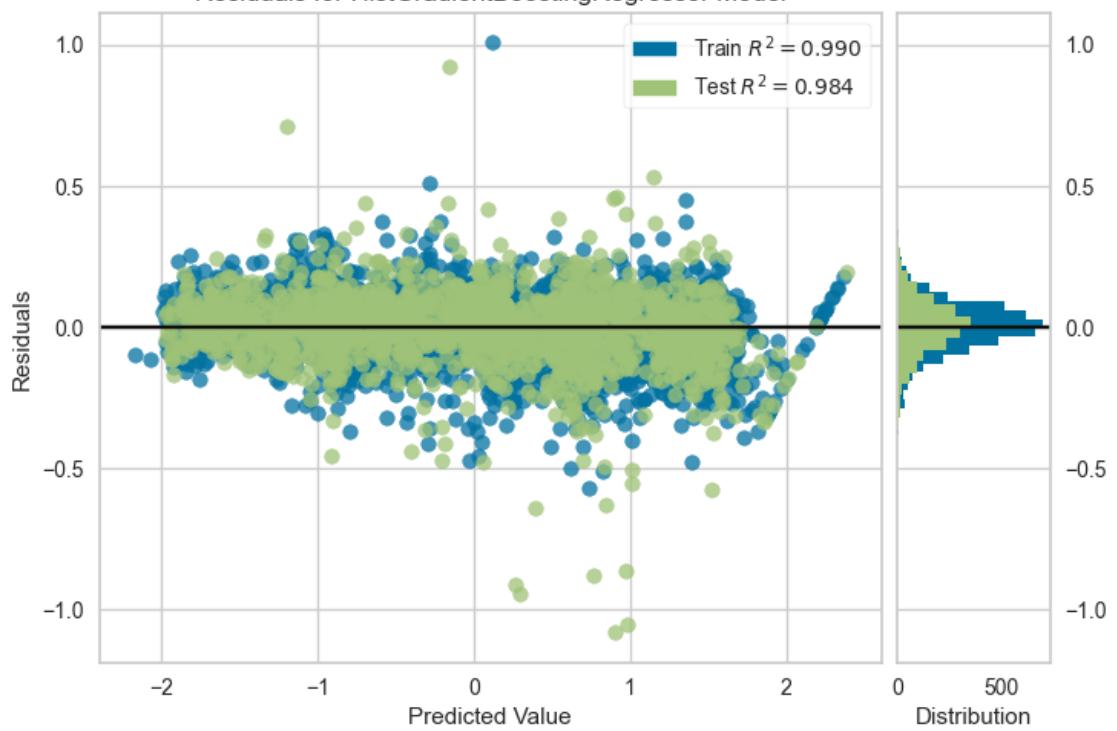


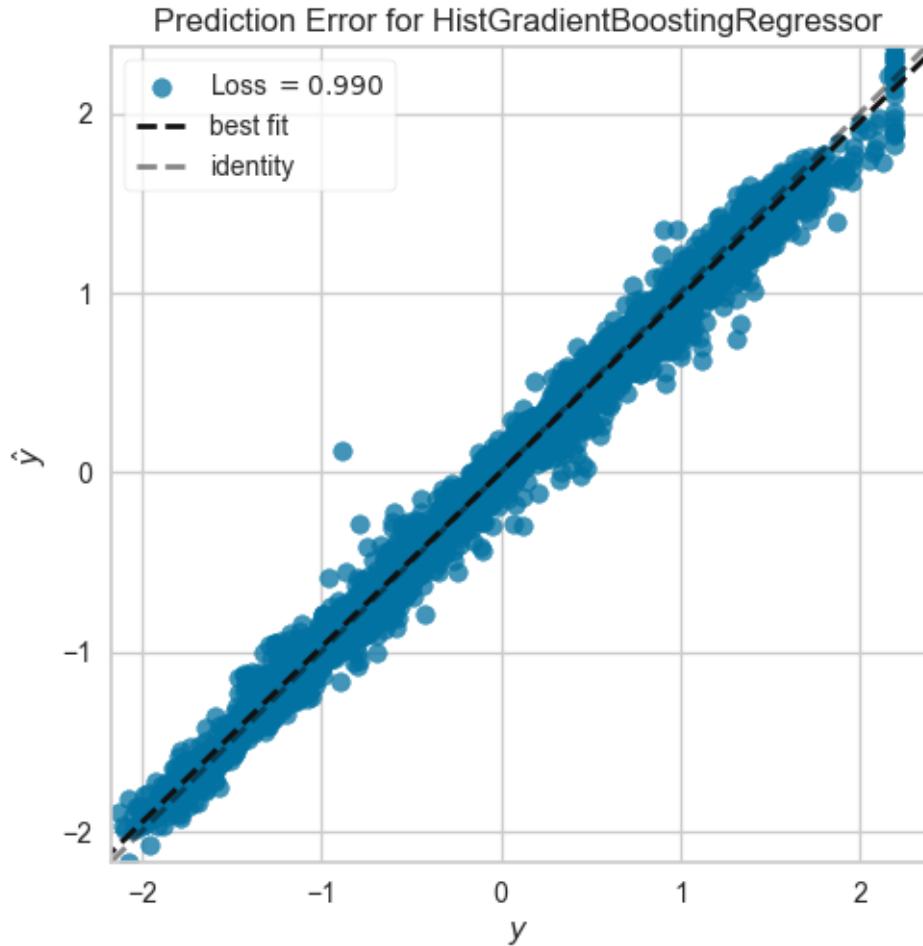


```
[79]: train_and_evaluate_model(HistGradientBoostingRegressor())
```

```
Mean Absolute Percentage Error(MAPE): 0.51087336067734
Root Mean Squared Error(RMSE): 0.1294334736270051
R2 Score: 0.9836342525411399
```

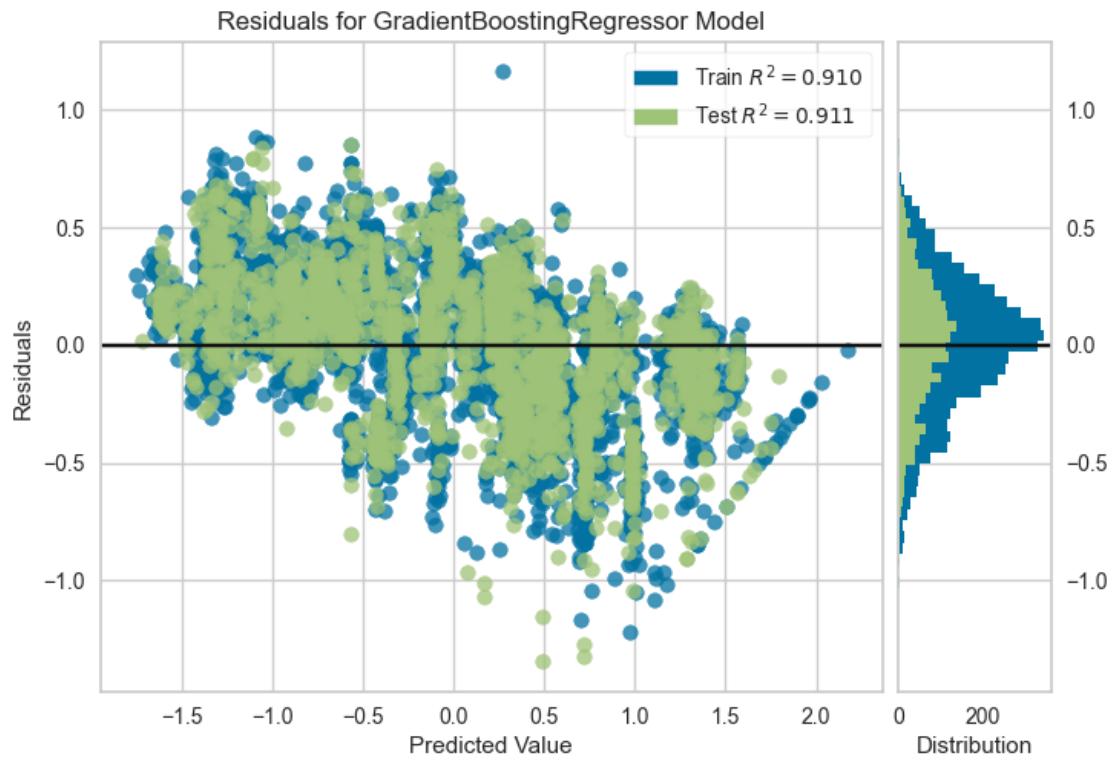
Residuals for HistGradientBoostingRegressor Model

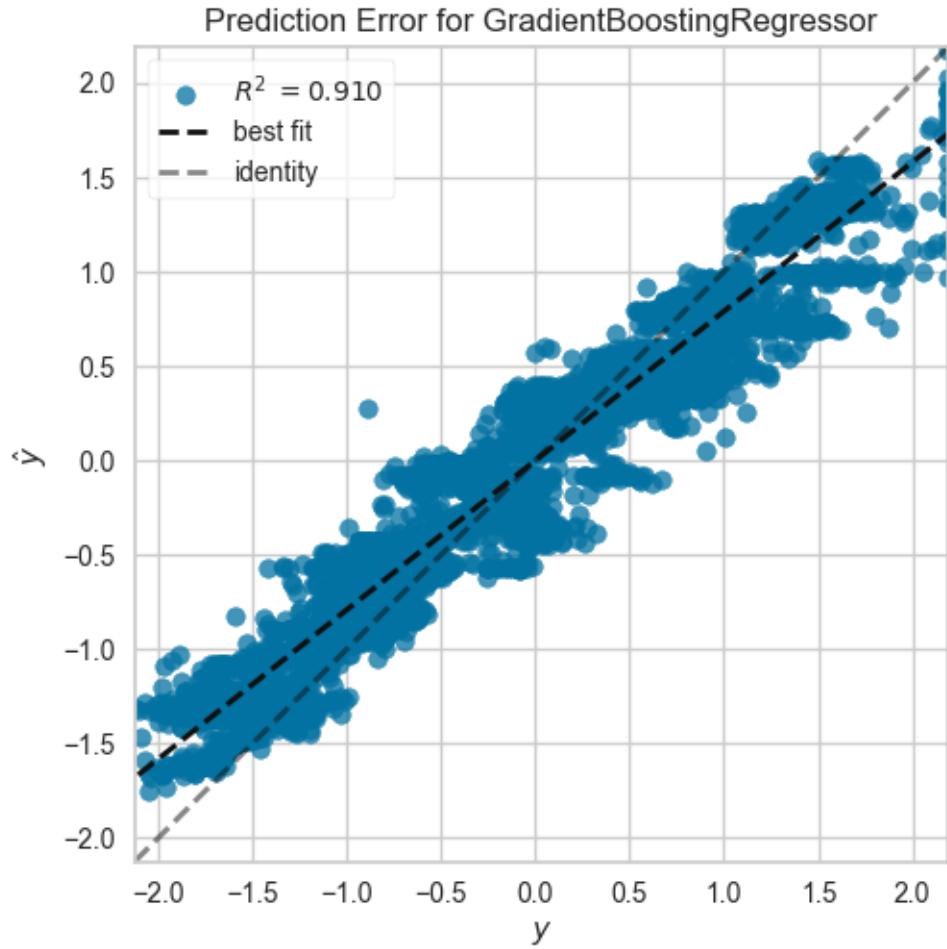




```
[80]: train_and_evaluate_model(GradientBoostingRegressor())
```

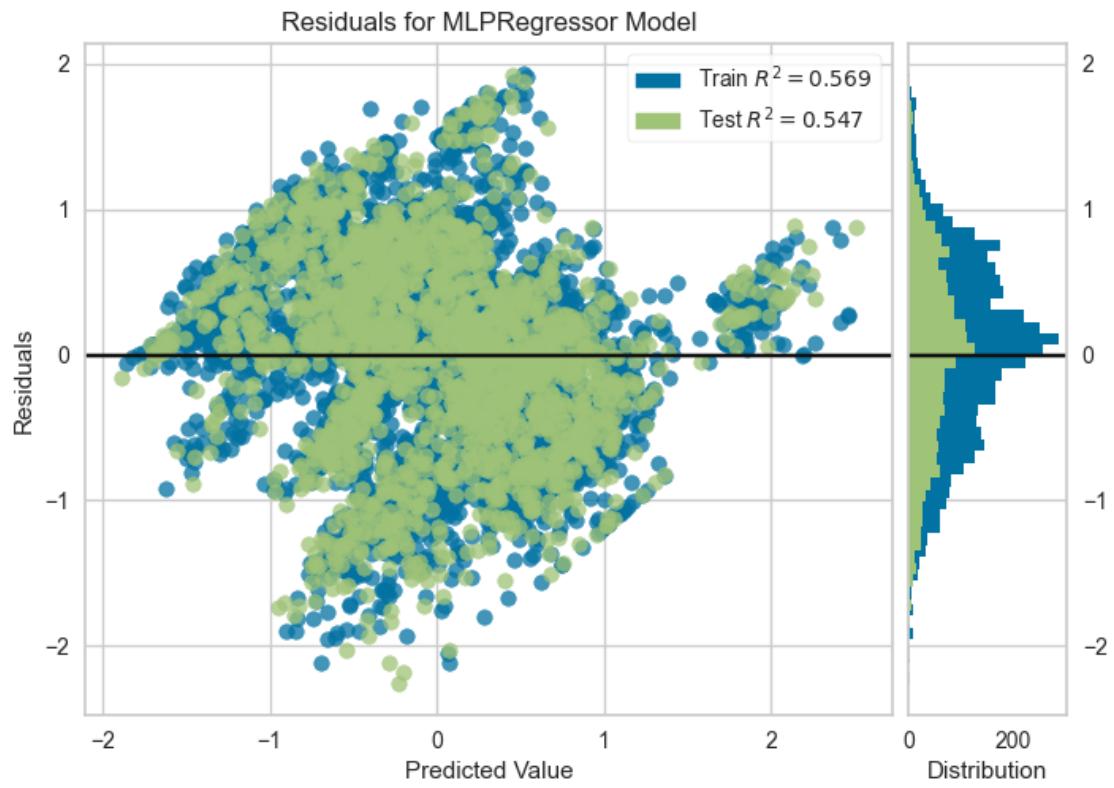
```
Mean Absolute Percentage Error(MAPE): 1.5252582037997267
Root Mean Squared Error(RMSE): 0.30209150905999316
R2 Score: 0.910850344465775
```

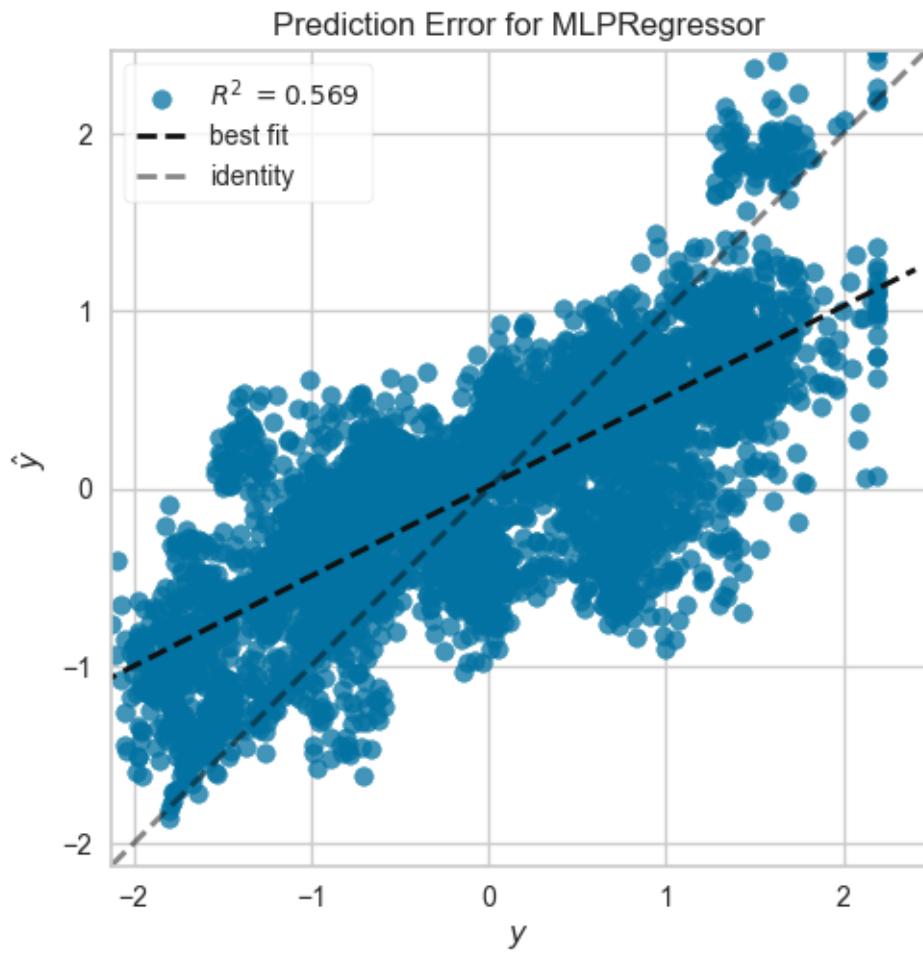




```
[81]: train_and_evaluate_model(MLPRegressor())
```

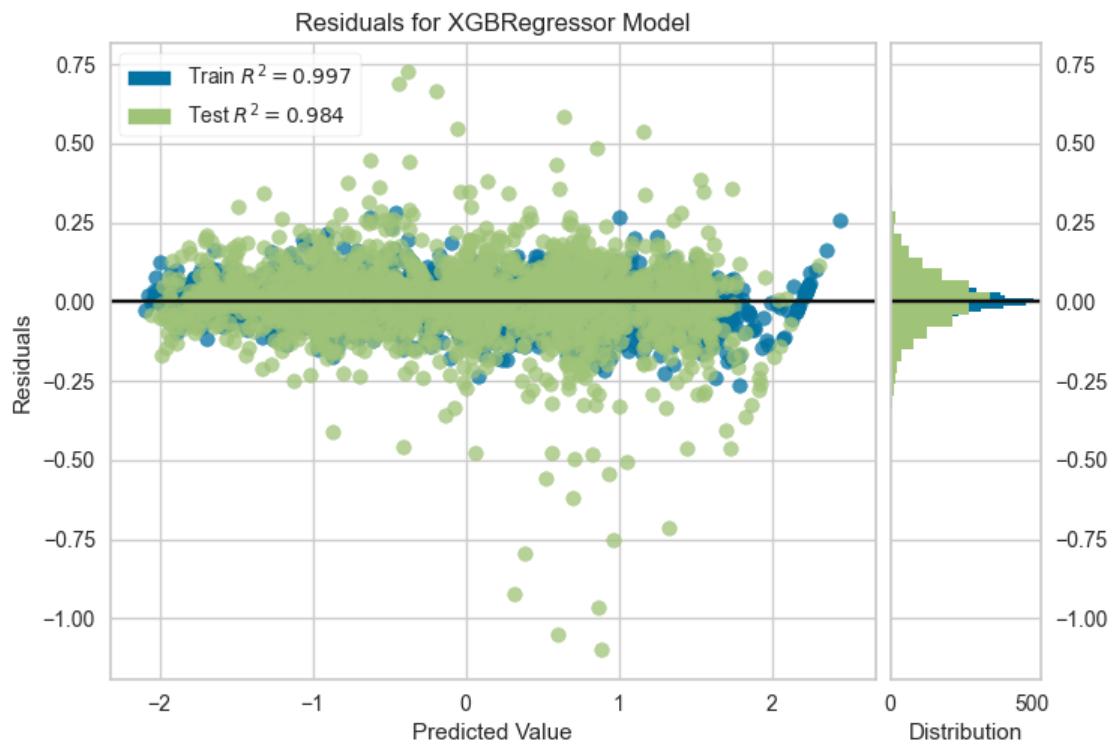
```
Mean Absolute Percentage Error(MAPE): 2.302538030869336
Root Mean Squared Error(RMSE): 0.6807573916414449
R2 Score: 0.5472824159249126
```

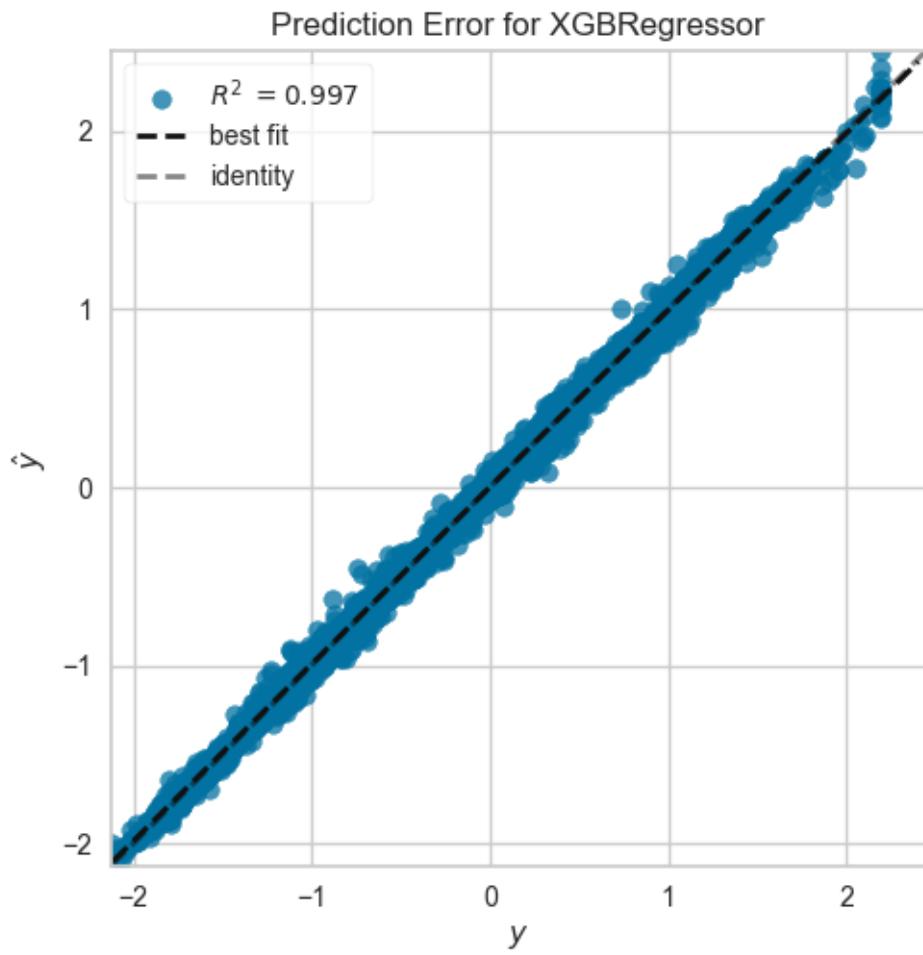




```
[82]: train_and_evaluate_model(XGBRegressor())
```

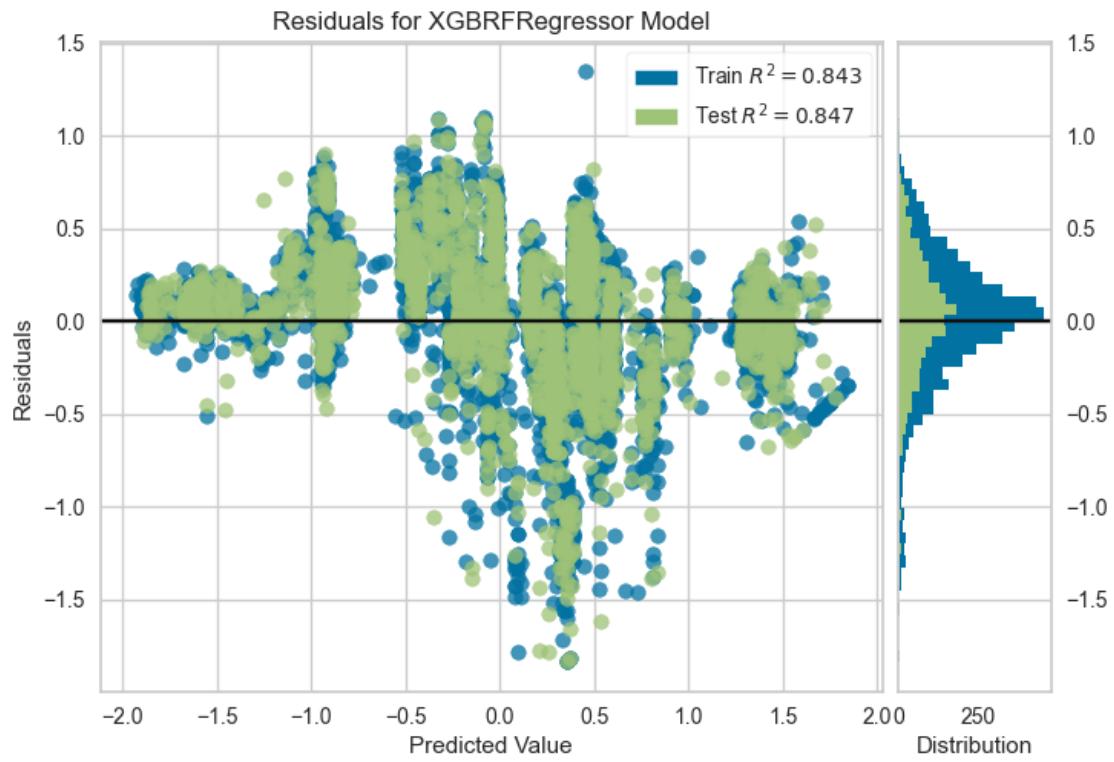
```
Mean Absolute Percentage Error(MAPE): 0.4484239317276413
Root Mean Squared Error(RMSE): 0.1279829255613287
R2 Score: 0.9839990156994279
```

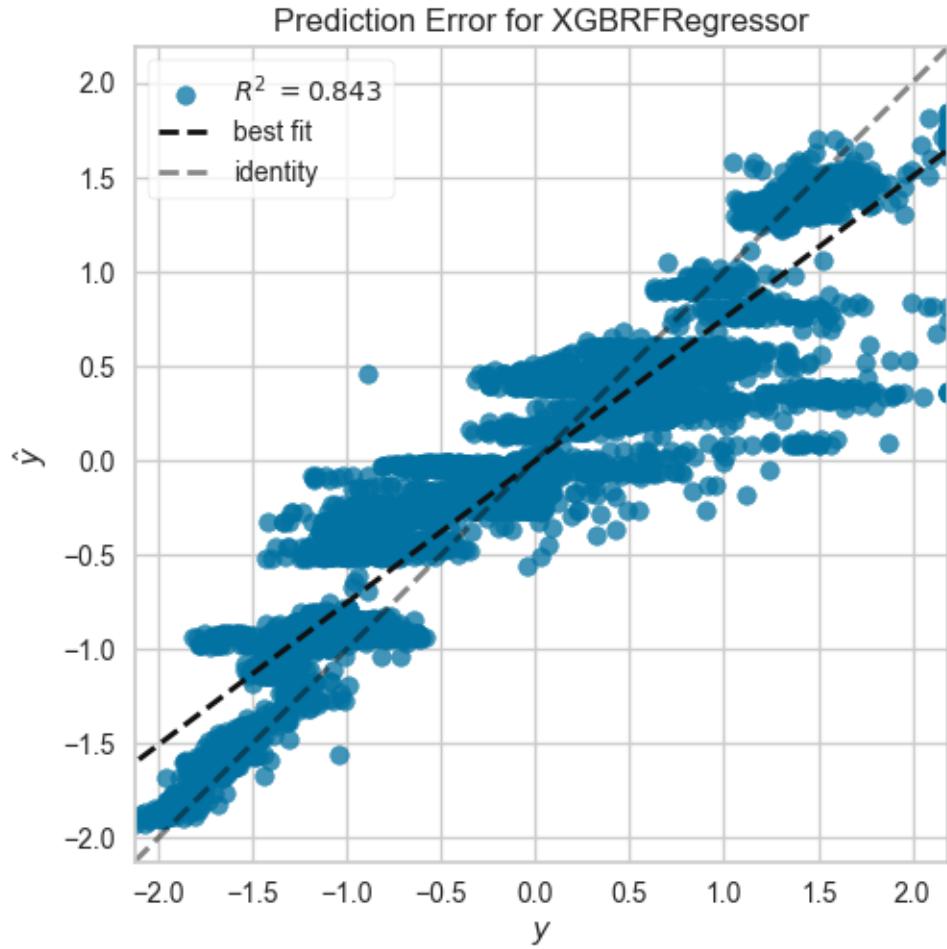




```
[83]: train_and_evaluate_model(XGBRFRegressor())
```

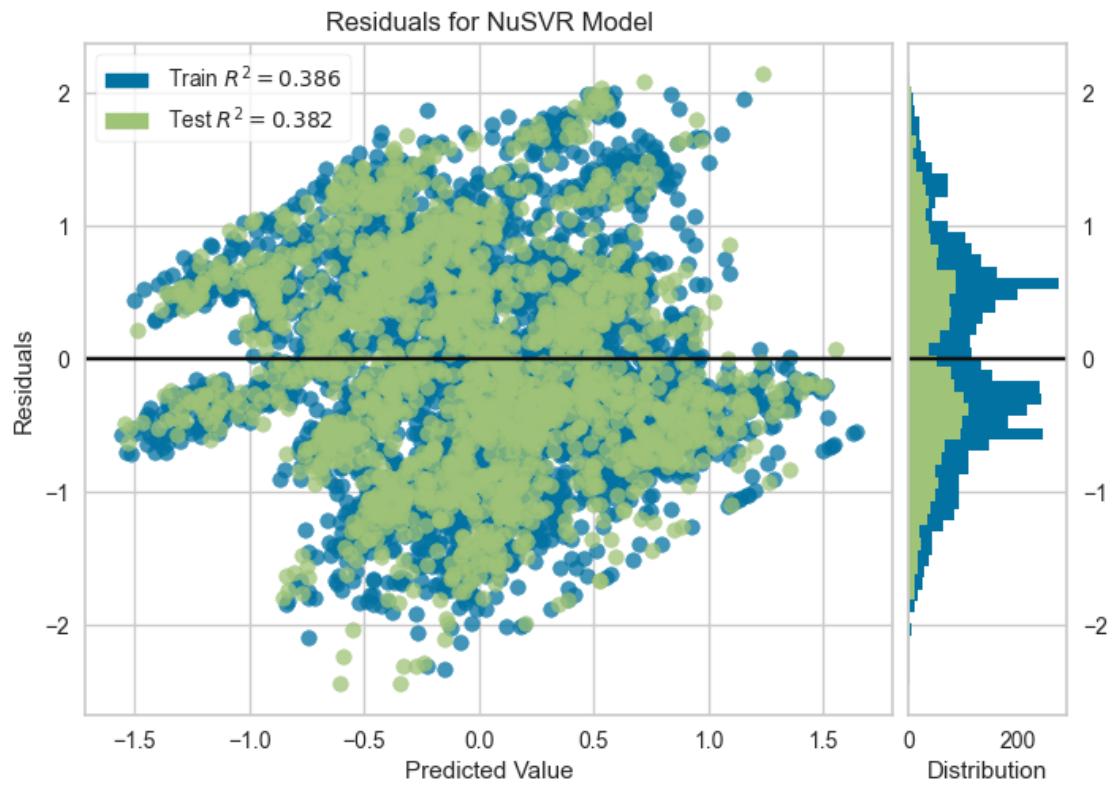
```
Mean Absolute Percentage Error(MAPE): 1.4958698963663275
Root Mean Squared Error(RMSE): 0.3956003717003603
R2 Score: 0.8471181201865408
```

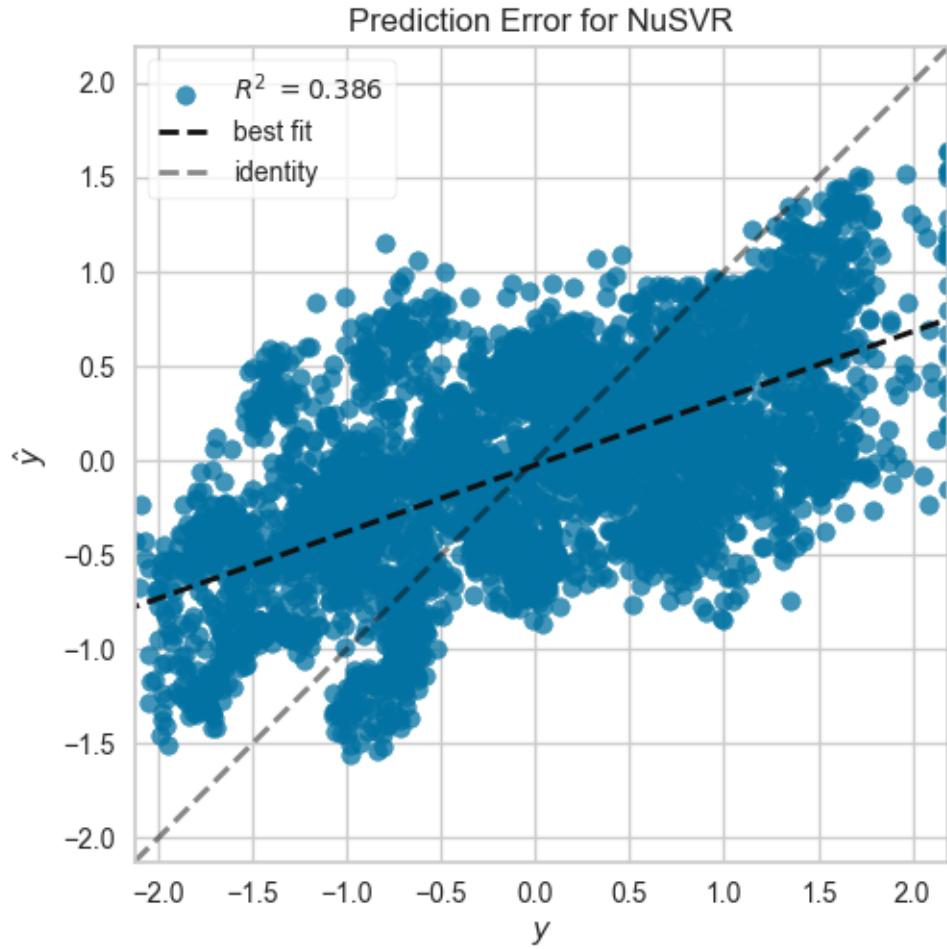




```
[84]: train_and_evaluate_model(NuSVR())
```

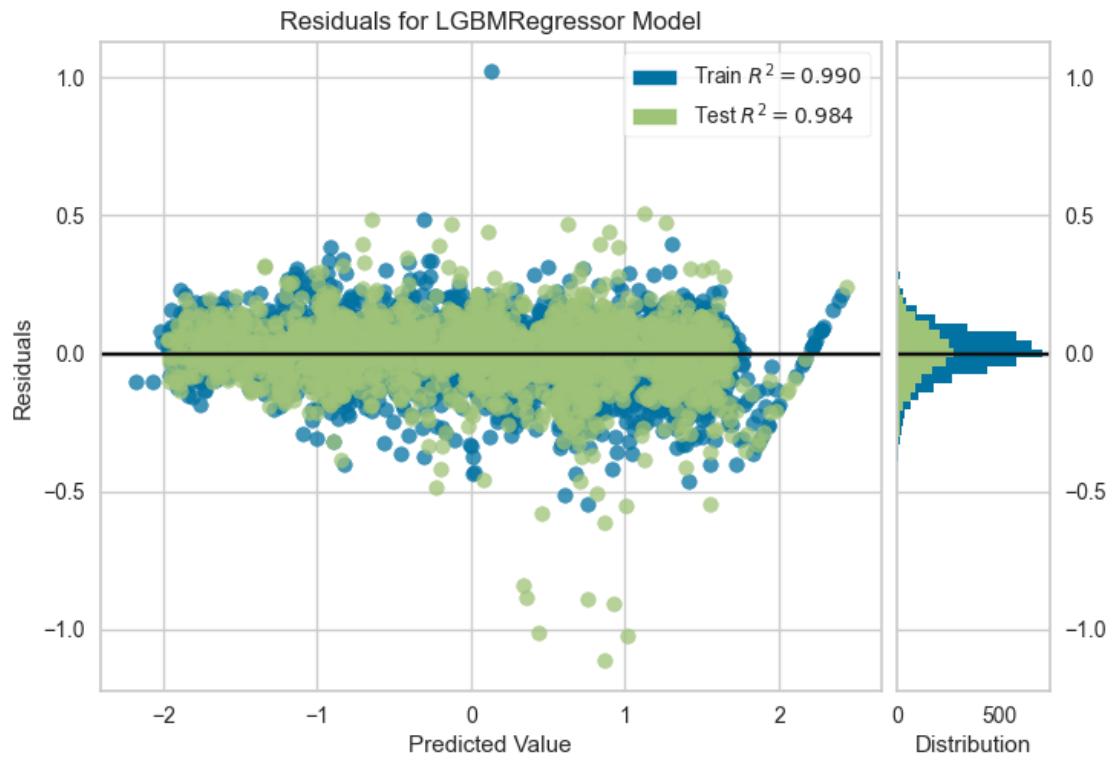
```
Mean Absolute Percentage Error(MAPE): 2.7676772849788933
Root Mean Squared Error(RMSE): 0.7953092639783247
R2 Score: 0.3821049519963209
```

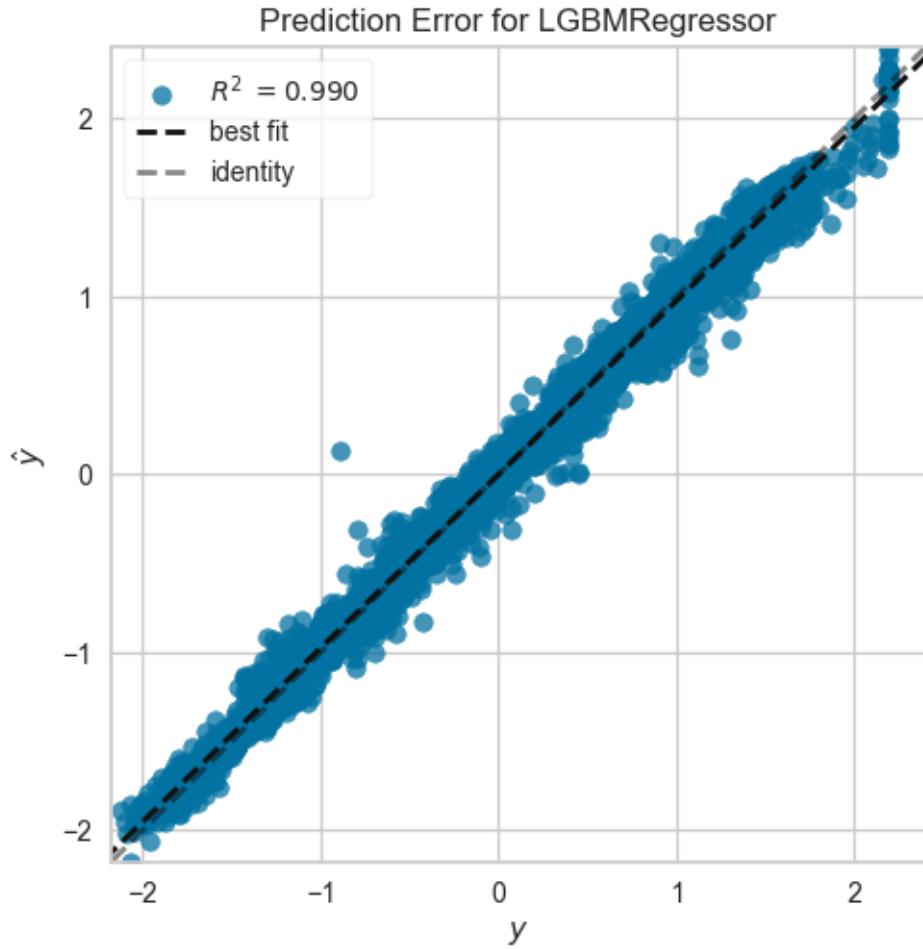




```
[85]: train_and_evaluate_model(LGBMRegressor())
```

```
Mean Absolute Percentage Error(MAPE): 0.4836466164864434
Root Mean Squared Error(RMSE): 0.1289760326076023
R2 Score: 0.9837497270767015
```





```
[86]: train_and_evaluate_model(CatBoostRegressor(silent=True))
```

Mean Absolute Percentage Error(MAPE): 0.5341453587985909  
Root Mean Squared Error(RMSE): 0.12290887063346671  
R2 Score: 0.9852426258469904

```
[145]: train_and_evaluate_model(VotingRegressor(estimators=[  

    ('CAT',CatBoostRegressor(silent=True)),  

    ('XGB',XGBRegressor()),  

    ('LGBM',LGBMRegressor()),  

    ('HGB',HistGradientBoostingRegressor()),  

    ('RF',RandomForestRegressor()),  

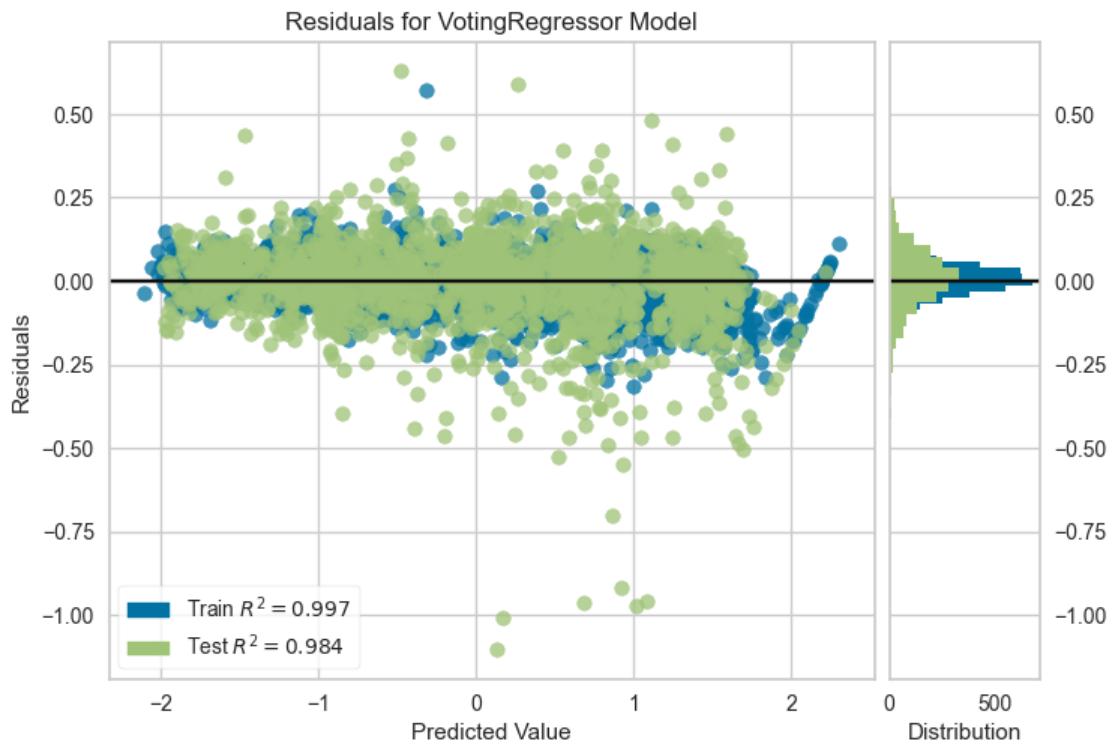
    ('BAG',BaggingRegressor()),  

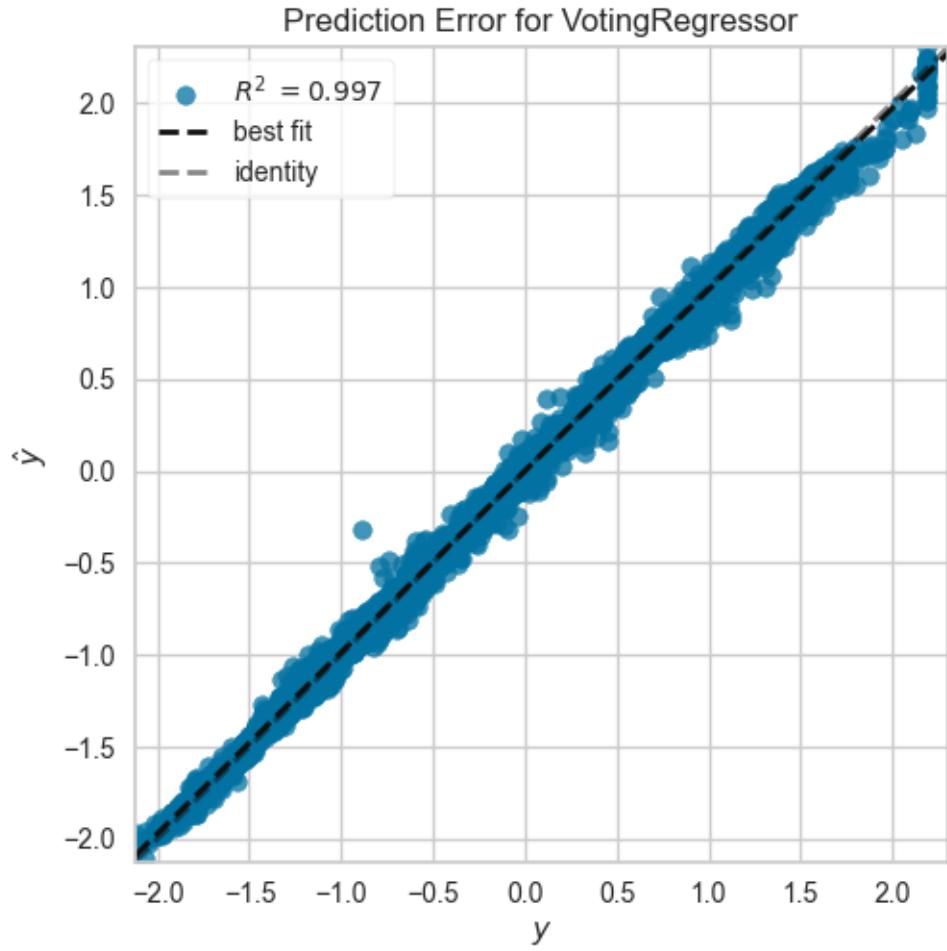
    ('ET',ExtraTreesRegressor())  

]))
```

Mean Absolute Percentage Error(MAPE): 0.4845606374418606  
Root Mean Squared Error(RMSE): 0.12994806863698197

R2 Score: 0.9835038616447688





## 0.7 Baseline Models Performance Comparison

```
[88]: model_perfs = pd.DataFrame({'Model': models, 'MAPE': mape_scores, 'RMSE': rmse_scores, 'R2': r2_scores})
model_perfs = model_perfs.sort_values('R2', ascending=False)
model_perfs
```

	Model	MAPE	RMSE	\
25	<catboost.core.CatBoostRegressor object at 0x0...>	0.534145	0.122909	
21	XGBRegressor(base_score=None, booster=None, ca...	0.448424	0.127983	
24	LGBMRegressor()	0.483647	0.128976	
18	HistGradientBoostingRegressor()	0.510873	0.129433	
26	VotingRegressor(estimators=[('CAT', \n          ...),\n          (ExtraTreeRegressor(random_state=1040630546), ...),\n          (DecisionTreeRegressor(max_features=1.0, rando...),\n          (DecisionTreeRegressor(random_state=1012004194...),\n          DecisionTreeRegressor())])	0.483340	0.131280	
16		0.596846	0.178494	
14		0.631636	0.180889	
15		0.628884	0.181817	
13		0.688617	0.229284	

```

19 ([DecisionTreeRegressor(criterion='friedman_ms...' 1.525258 0.302092
22 XGBRFRegressor(base_score=None, booster=None, ... 1.495870 0.395600
20                                         MLPRegressor() 2.302538 0.680757
17 (DecisionTreeRegressor(max_depth=3, random_st... 2.575329 0.725043
10                                         KNeighborsRegressor() 1.832426 0.725657
23                                         NuSVR() 2.767677 0.795309
11                                         SVR() 1.778644 0.815938
0                                         LinearRegression() 1.858860 0.942806
2                                         RidgeCV() 1.858337 0.942913
4                                         SGDRegressor() 1.824808 0.942968
1                                         LassoCV() 1.872174 0.943096
3                                         ElasticNetCV() 1.869417 0.943112
6                                         ARDRegression() 1.896284 0.943125
5                                         HuberRegressor() 2.047510 0.945681
12                                         LinearSVR() 2.058692 0.953704
8                                         TweedieRegressor() 1.300327 0.964521
9                                         PassiveAggressiveRegressor() 5.203633 1.123522
7                                         RANSACRegressor() 4.191927 1.257529

```

	R2
25	0.985243
21	0.983999
24	0.983750
18	0.983634
26	0.983164
16	0.968876
14	0.968036
15	0.967707
13	0.948644
19	0.910850
22	0.847118
20	0.547282
17	0.486465
10	0.485594
23	0.382105
11	0.349636
0	0.131666
2	0.131468
4	0.131367
1	0.131130
3	0.131102
6	0.131078
5	0.126361
12	0.111474
8	0.091205
9	-0.233122
7	-0.544823

Among the baseline models, Cat Boost Regressor produced the best performance by obtaining an astonishing r2 score of more than 98.5%.

## 0.8 Hyperparameter Optimization and Cross Validation

```
[89]: param_grid = {'n_estimators': [200,400,600,800,1000],
                  'criterion': ['squared_error', 'absolute_error', 'friedman_mse', 'poisson'],
                  'max_features': ['auto','sqrt','log2'],
                  'bootstrap': [True,False],
                  'oob_score': [True,False]}

grid_rf = RandomizedSearchCV(RandomForestRegressor(),param_grid,verbose=2,cv=5)
train_and_evaluate_model(grid_rf)
```

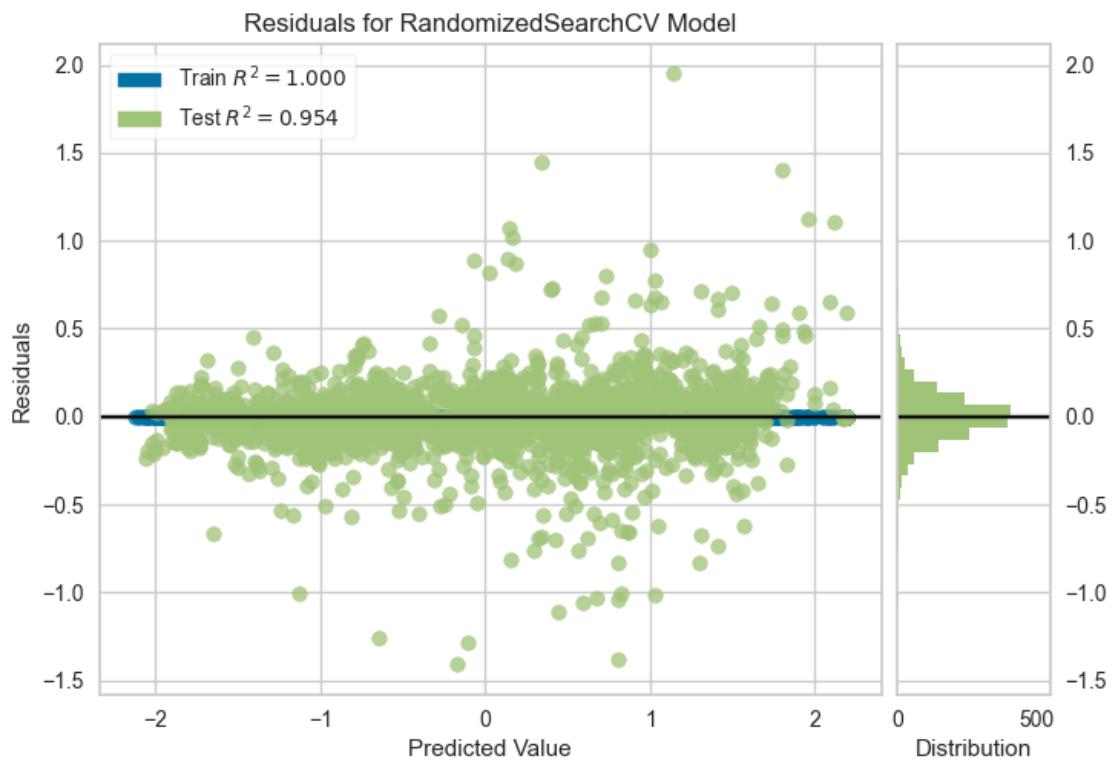
```
Fitting 5 folds for each of 10 candidates, totalling 50 fits
[CV] END bootstrap=True, criterion=friedman_mse, max_features=log2,
n_estimators=600, oob_score=True; total time= 5.2s
[CV] END bootstrap=True, criterion=friedman_mse, max_features=log2,
n_estimators=600, oob_score=True; total time= 5.1s
[CV] END bootstrap=True, criterion=friedman_mse, max_features=log2,
n_estimators=600, oob_score=True; total time= 5.2s
[CV] END bootstrap=True, criterion=friedman_mse, max_features=log2,
n_estimators=600, oob_score=True; total time= 5.2s
[CV] END bootstrap=True, criterion=friedman_mse, max_features=log2,
n_estimators=600, oob_score=True; total time= 6.0s
[CV] END bootstrap=True, criterion=absolute_error, max_features=log2,
n_estimators=200, oob_score=False; total time= 12.6s
[CV] END bootstrap=True, criterion=absolute_error, max_features=log2,
n_estimators=200, oob_score=False; total time= 12.8s
[CV] END bootstrap=True, criterion=absolute_error, max_features=log2,
n_estimators=200, oob_score=False; total time= 12.9s
[CV] END bootstrap=True, criterion=absolute_error, max_features=log2,
n_estimators=200, oob_score=False; total time= 12.5s
[CV] END bootstrap=True, criterion=absolute_error, max_features=log2,
n_estimators=200, oob_score=False; total time= 14.5s
[CV] END bootstrap=False, criterion=poisson, max_features=log2,
n_estimators=200, oob_score=False; total time= 0.0s
[CV] END bootstrap=False, criterion=poisson, max_features=log2,
n_estimators=200, oob_score=False; total time= 0.0s
[CV] END bootstrap=False, criterion=poisson, max_features=log2,
n_estimators=200, oob_score=False; total time= 0.0s
[CV] END bootstrap=False, criterion=poisson, max_features=log2,
n_estimators=200, oob_score=False; total time= 0.0s
[CV] END bootstrap=False, criterion=poisson, max_features=log2,
n_estimators=200, oob_score=False; total time= 0.0s
[CV] END bootstrap=True, criterion=poisson, max_features=log2, n_estimators=400,
oob_score=True; total time= 0.0s
```

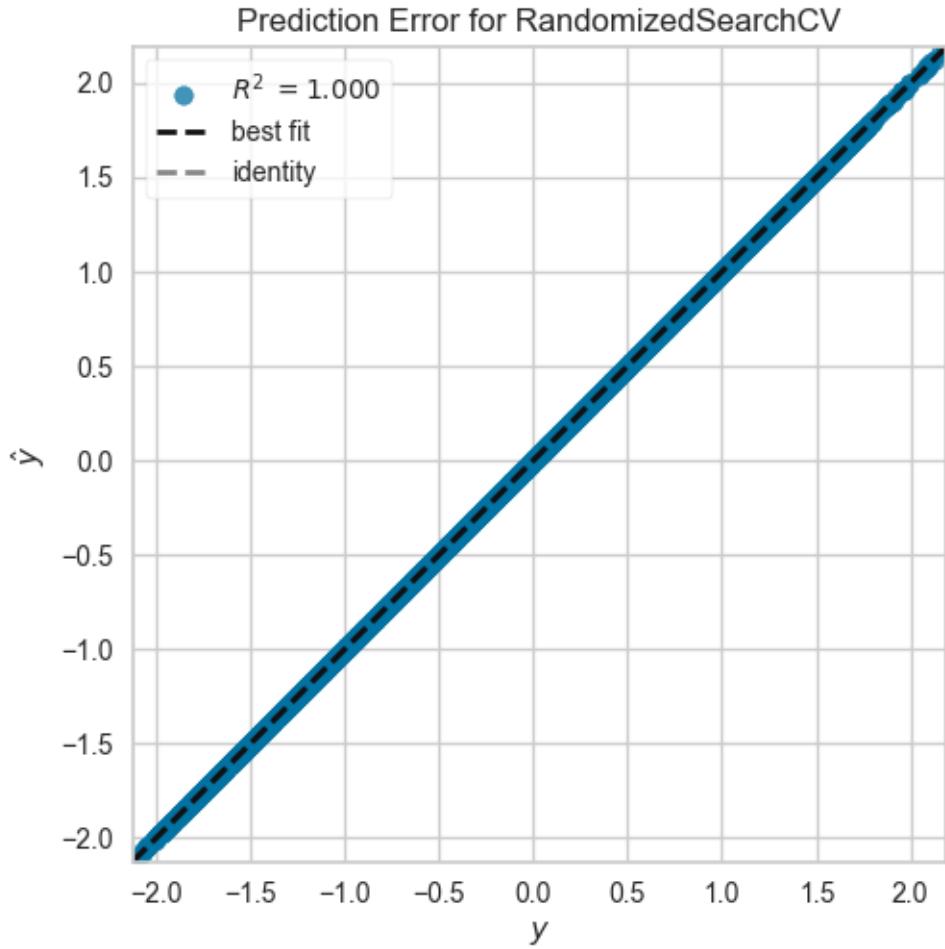
```
[CV] END bootstrap=True, criterion=poisson, max_features=log2, n_estimators=400,
oob_score=True; total time= 0.0s
[CV] END bootstrap=True, criterion=poisson, max_features=log2, n_estimators=400,
oob_score=True; total time= 0.0s
[CV] END bootstrap=True, criterion=poisson, max_features=log2, n_estimators=400,
oob_score=True; total time= 0.0s
[CV] END bootstrap=True, criterion=poisson, max_features=log2, n_estimators=400,
oob_score=True; total time= 0.0s
[CV] END bootstrap=False, criterion=friedman_mse, max_features=auto,
n_estimators=1000, oob_score=False; total time= 18.2s
[CV] END bootstrap=False, criterion=friedman_mse, max_features=auto,
n_estimators=1000, oob_score=False; total time= 18.0s
[CV] END bootstrap=False, criterion=friedman_mse, max_features=auto,
n_estimators=1000, oob_score=False; total time= 18.0s
[CV] END bootstrap=False, criterion=friedman_mse, max_features=auto,
n_estimators=1000, oob_score=False; total time= 19.0s
[CV] END bootstrap=False, criterion=friedman_mse, max_features=auto,
n_estimators=1000, oob_score=False; total time= 18.2s
[CV] END bootstrap=False, criterion=absolute_error, max_features=auto,
n_estimators=400, oob_score=True; total time= 0.0s
[CV] END bootstrap=False, criterion=absolute_error, max_features=auto,
n_estimators=400, oob_score=True; total time= 0.0s
[CV] END bootstrap=False, criterion=absolute_error, max_features=auto,
n_estimators=400, oob_score=True; total time= 0.0s
[CV] END bootstrap=False, criterion=absolute_error, max_features=auto,
n_estimators=400, oob_score=True; total time= 0.0s
[CV] END bootstrap=False, criterion=absolute_error, max_features=auto,
n_estimators=400, oob_score=True; total time= 0.0s
[CV] END bootstrap=True, criterion=squared_error, max_features=log2,
n_estimators=800, oob_score=True; total time= 7.1s
[CV] END bootstrap=True, criterion=squared_error, max_features=log2,
n_estimators=800, oob_score=True; total time= 7.1s
[CV] END bootstrap=True, criterion=squared_error, max_features=log2,
n_estimators=800, oob_score=True; total time= 7.0s
[CV] END bootstrap=True, criterion=squared_error, max_features=log2,
n_estimators=800, oob_score=True; total time= 7.0s
[CV] END bootstrap=True, criterion=squared_error, max_features=log2,
n_estimators=800, oob_score=True; total time= 7.1s
[CV] END bootstrap=False, criterion=squared_error, max_features=auto,
n_estimators=400, oob_score=False; total time= 7.9s
[CV] END bootstrap=False, criterion=squared_error, max_features=auto,
n_estimators=400, oob_score=False; total time= 7.0s
[CV] END bootstrap=False, criterion=squared_error, max_features=auto,
n_estimators=400, oob_score=False; total time= 7.0s
[CV] END bootstrap=False, criterion=squared_error, max_features=auto,
n_estimators=400, oob_score=False; total time= 7.1s
[CV] END bootstrap=False, criterion=squared_error, max_features=auto,
n_estimators=400, oob_score=False; total time= 7.1s
```

```

[CV] END bootstrap=False, criterion=absolute_error, max_features=sqrt,
n_estimators=1000, oob_score=True; total time= 0.0s
[CV] END bootstrap=False, criterion=absolute_error, max_features=sqrt,
n_estimators=1000, oob_score=True; total time= 0.0s
[CV] END bootstrap=False, criterion=absolute_error, max_features=sqrt,
n_estimators=1000, oob_score=True; total time= 0.0s
[CV] END bootstrap=False, criterion=absolute_error, max_features=sqrt,
n_estimators=1000, oob_score=True; total time= 0.0s
[CV] END bootstrap=False, criterion=absolute_error, max_features=sqrt,
n_estimators=1000, oob_score=True; total time= 0.0s
[CV] END bootstrap=False, criterion=absolute_error, max_features=sqrt,
n_estimators=1000, oob_score=True; total time= 0.0s
[CV] END bootstrap=True, criterion=squared_error, max_features=log2,
n_estimators=600, oob_score=True; total time= 5.2s
[CV] END bootstrap=True, criterion=squared_error, max_features=log2,
n_estimators=600, oob_score=True; total time= 5.3s
[CV] END bootstrap=True, criterion=squared_error, max_features=log2,
n_estimators=600, oob_score=True; total time= 5.2s
[CV] END bootstrap=True, criterion=squared_error, max_features=log2,
n_estimators=600, oob_score=True; total time= 5.3s
[CV] END bootstrap=True, criterion=squared_error, max_features=log2,
n_estimators=600, oob_score=True; total time= 5.3s
[CV] END bootstrap=True, criterion=squared_error, max_features=log2,
n_estimators=600, oob_score=True; total time= 5.3s
Mean Absolute Percentage Error(MAPE): 0.6704343334863478
Root Mean Squared Error(RMSE): 0.21668601661566092
R2 Score: 0.9541325703008455

```





```
[90]: grid_rf.best_params_
```

```
[90]: {'oob_score': False,
       'n_estimators': 400,
       'max_features': 'auto',
       'criterion': 'squared_error',
       'bootstrap': False}
```

```
[91]: param_grid = {'n_neighbors': [2,5,8,12,20],
                  'weights': ['uniform','distance'],
                  'algorithm': ['ball_tree', 'kd_tree', 'brute'],
                  'metric': ['minkowski','manhattan','euclidean','chebyshev']}
}

grid_knn = RandomizedSearchCV(KNeighborsRegressor(),param_grid, cv=RepeatedKFold(n_splits=5,n_repeats=3)
train_and_evaluate_model(grid_knn)
```









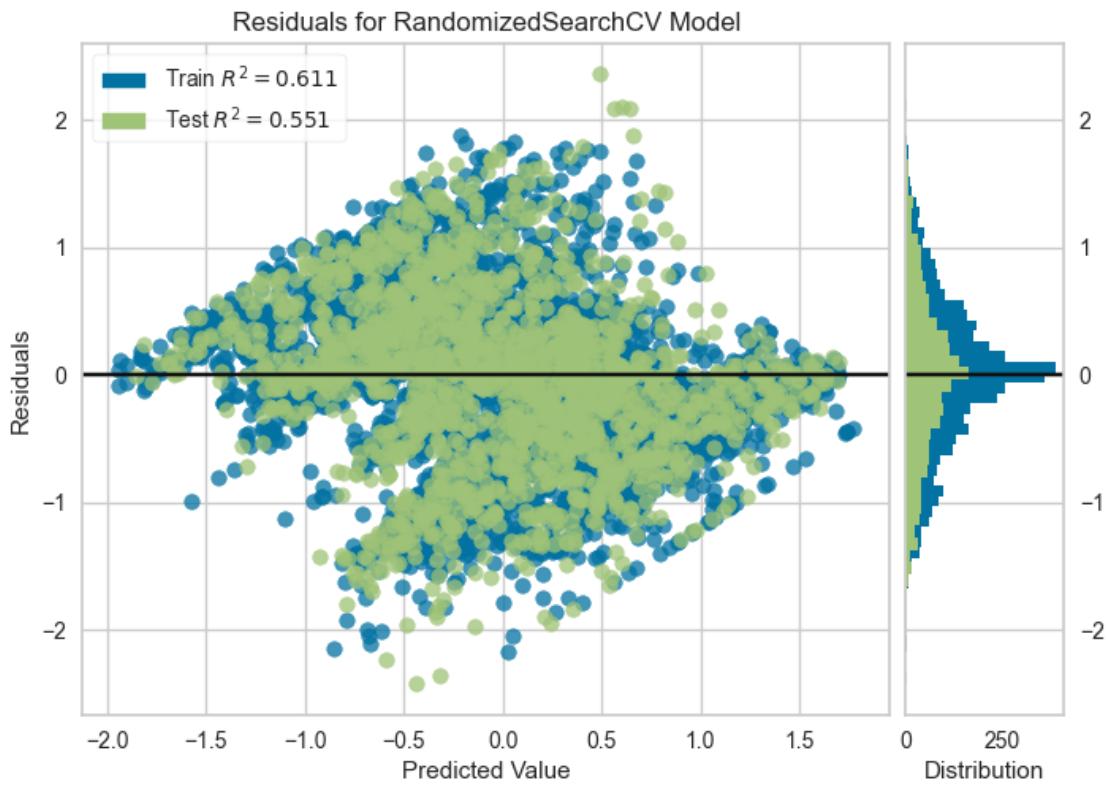


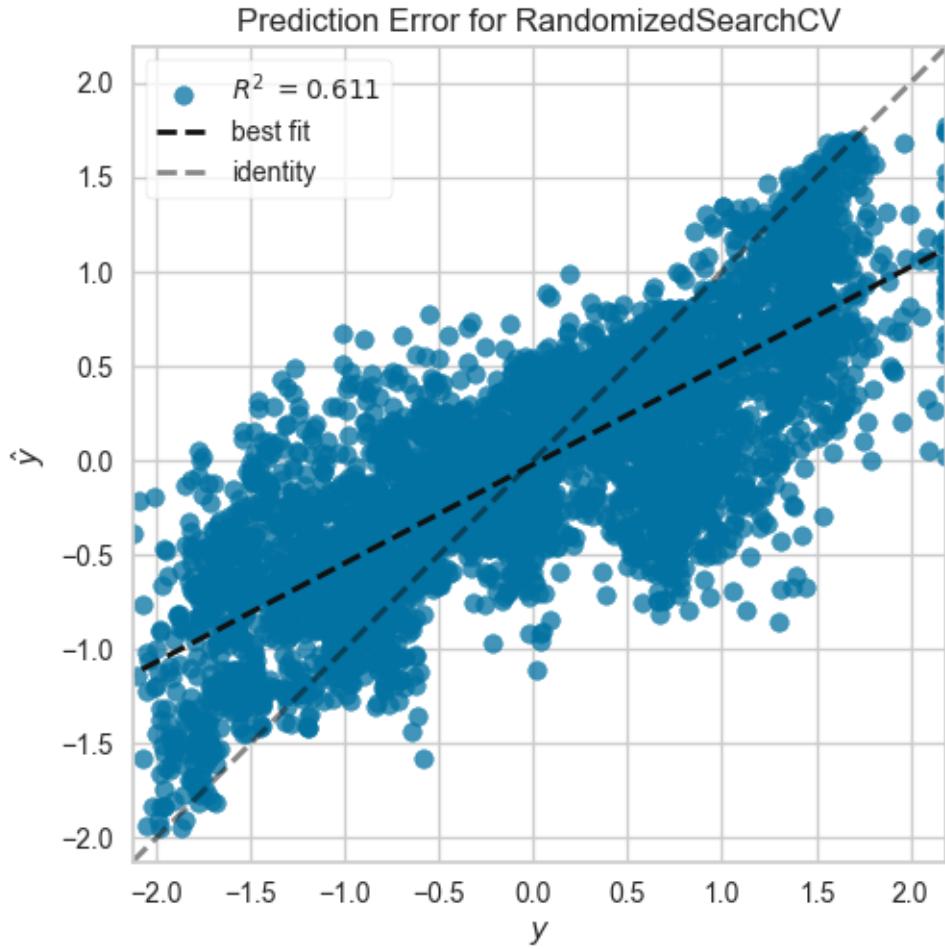


```

total time= 0.0s
[CV] END algorithm=kd_tree, metric=chebyshev, n_neighbors=20, weights=distance;
total time= 0.0s
[CV] END algorithm=kd_tree, metric=chebyshev, n_neighbors=20, weights=distance;
total time= 0.0s
[CV] END algorithm=kd_tree, metric=chebyshev, n_neighbors=20, weights=distance;
total time= 0.0s
[CV] END algorithm=kd_tree, metric=chebyshev, n_neighbors=20, weights=distance;
total time= 0.0s
[CV] END algorithm=kd_tree, metric=chebyshev, n_neighbors=20, weights=distance;
total time= 0.0s
[CV] END algorithm=kd_tree, metric=chebyshev, n_neighbors=20, weights=distance;
total time= 0.0s
Mean Absolute Percentage Error(MAPE): 1.8270423271050913
Root Mean Squared Error(RMSE): 0.6781109353399526
R2 Score: 0.5507954691967909

```





```
[92]: grid_knn.best_params_
```

```
[92]: {'weights': 'uniform',
       'n_neighbors': 12,
       'metric': 'manhattan',
       'algorithm': 'ball_tree'}
```

```
[93]: param_grid = {'learning_rate': [0.2, 0.4, 0.5, 0.8, 1.0],
                  'loss': ['squared_error', 'absolute_error', 'poisson', 'quantile'],
                  'max_bins': np.arange(0, 255, 50),
                  'interaction_cst': ['pairwise', 'no_interaction']}
grid_hgb = RandomizedSearchCV(HistGradientBoostingRegressor(), param_grid, cv=RepeatedKFold(n_splits=5, n_repeats=3))
train_and_evaluate_model(grid_hgb)
```

Fitting 15 folds for each of 10 candidates, totalling 150 fits  
[CV] END interaction\_cst=pairwise, learning\_rate=0.8, loss=poisson, max\_bins=0;











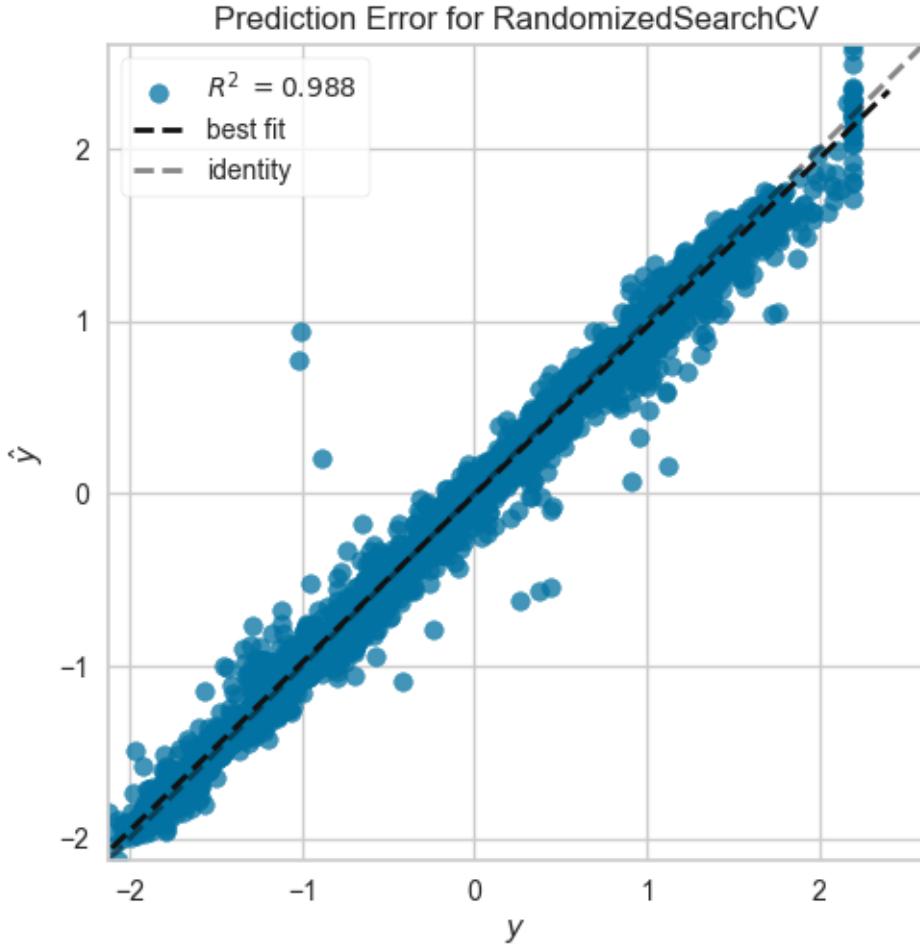


```

max_bins=250; total time= 0.0s
[CV] END interaction_cst=pairwise, learning_rate=0.2, loss=quantile,
max_bins=250; total time= 0.0s
[CV] END interaction_cst=pairwise, learning_rate=0.2, loss=quantile,
max_bins=250; total time= 0.0s
[CV] END interaction_cst=pairwise, learning_rate=0.2, loss=quantile,
max_bins=250; total time= 0.0s
[CV] END interaction_cst=pairwise, learning_rate=0.2, loss=quantile,
max_bins=250; total time= 0.0s
[CV] END interaction_cst=pairwise, learning_rate=0.2, loss=quantile,
max_bins=250; total time= 0.0s
Mean Absolute Percentage Error(MAPE): 0.353993990408274
Root Mean Squared Error(RMSE): 0.1286878540246637
R2 Score: 0.9838222637922769

```





```
[94]: grid_hgb.best_params_
```

```
[94]: {'max_bins': 200,
       'loss': 'absolute_error',
       'learning_rate': 0.2,
       'interaction_cst': 'pairwise'}
```

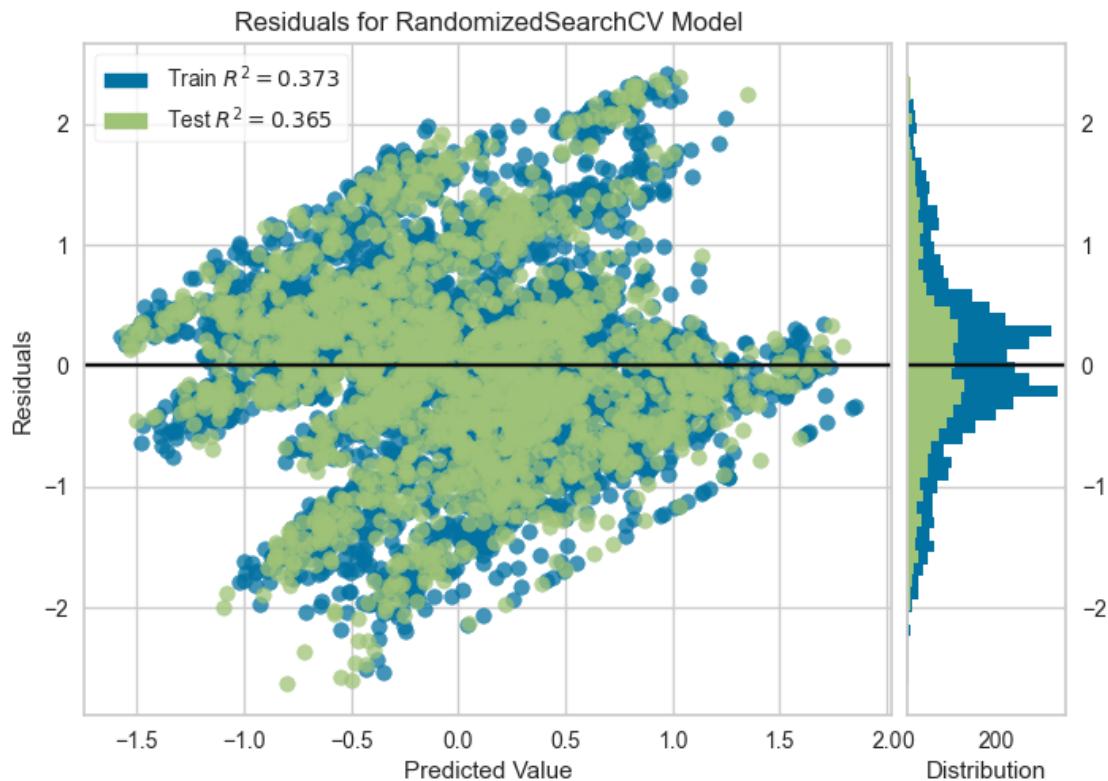
```
[95]: param_grid = {'C': [0.001, 0.01, 0.1, 1, 5],
                  'gamma': ['scale', 'auto'],
                  'epsilon': np.linspace(0.001, 1, 5),
                  'kernel': ['linear', 'poly', 'rbf', 'sigmoid'],
                  'degree': [2, 3, 4, 5],
                  'shrinking': [True, False]
                 }
grid_svr = RandomizedSearchCV(SVR(), param_grid, verbose=2, cv=5)
train_and_evaluate_model(grid_svr)
```

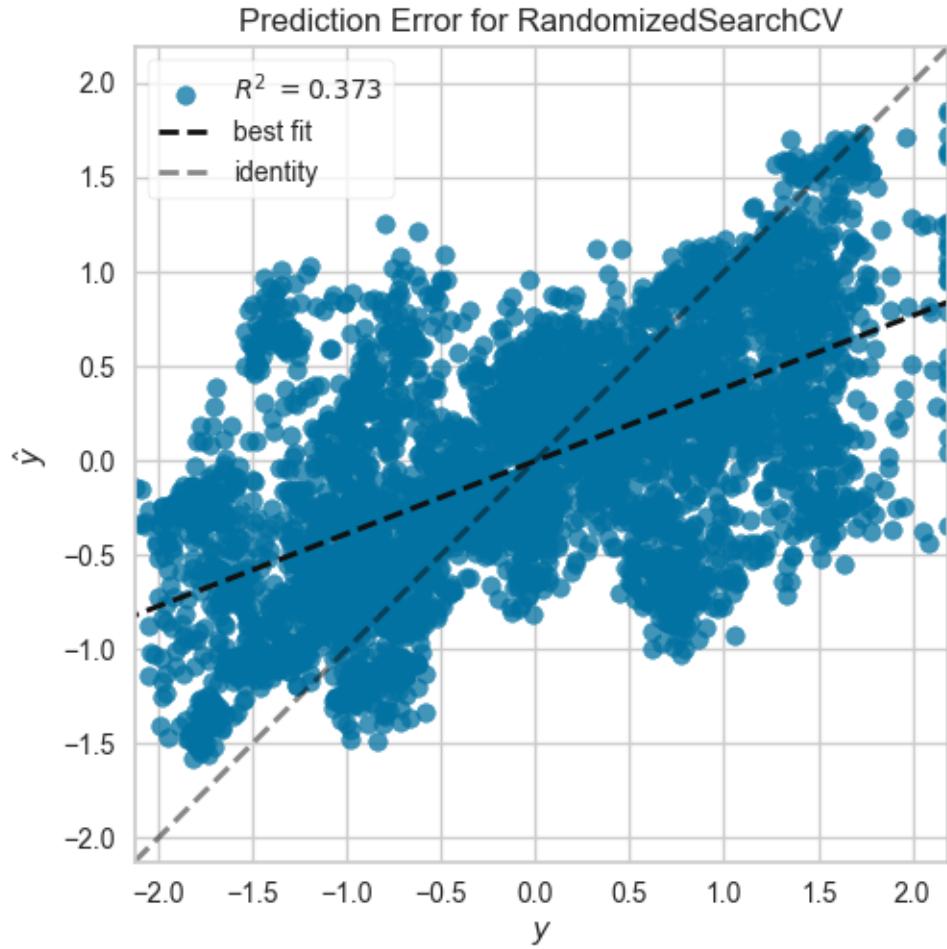
Fitting 5 folds for each of 10 candidates, totalling 50 fits

```
[CV] END C=0.1, degree=5, epsilon=1.0, gamma=scale, kernel=linear,  
shrinking=True; total time= 0.1s  
[CV] END C=0.1, degree=5, epsilon=1.0, gamma=scale, kernel=linear,  
shrinking=True; total time= 0.1s  
[CV] END C=0.1, degree=5, epsilon=1.0, gamma=scale, kernel=linear,  
shrinking=True; total time= 0.1s  
[CV] END C=0.1, degree=5, epsilon=1.0, gamma=scale, kernel=linear,  
shrinking=True; total time= 0.1s  
[CV] END C=0.1, degree=5, epsilon=1.0, gamma=scale, kernel=linear,  
shrinking=True; total time= 0.1s  
[CV] END C=1, degree=3, epsilon=0.25075, gamma=scale, kernel=rbf,  
shrinking=True; total time= 0.5s  
[CV] END C=1, degree=3, epsilon=0.25075, gamma=scale, kernel=rbf,  
shrinking=True; total time= 0.5s  
[CV] END C=1, degree=3, epsilon=0.25075, gamma=scale, kernel=rbf,  
shrinking=True; total time= 0.5s  
[CV] END C=1, degree=3, epsilon=0.25075, gamma=scale, kernel=rbf,  
shrinking=True; total time= 0.6s  
[CV] END C=1, degree=3, epsilon=0.25075, gamma=scale, kernel=rbf,  
shrinking=True; total time= 0.6s  
[CV] END C=0.1, degree=3, epsilon=0.5005, gamma=auto, kernel=linear,  
shrinking=True; total time= 0.2s  
[CV] END C=0.1, degree=3, epsilon=0.5005, gamma=auto, kernel=linear,  
shrinking=True; total time= 0.2s  
[CV] END C=0.1, degree=3, epsilon=0.5005, gamma=auto, kernel=linear,  
shrinking=True; total time= 0.4s  
[CV] END C=0.1, degree=3, epsilon=0.5005, gamma=auto, kernel=linear,  
shrinking=True; total time= 0.6s  
[CV] END C=0.1, degree=3, epsilon=0.5005, gamma=auto, kernel=linear,  
shrinking=True; total time= 0.6s  
[CV] END C=0.1, degree=4, epsilon=0.25075, gamma=scale, kernel=sigmoid,  
shrinking=False; total time= 0.7s  
[CV] END C=0.1, degree=4, epsilon=0.25075, gamma=scale, kernel=sigmoid,  
shrinking=False; total time= 0.7s  
[CV] END C=0.1, degree=4, epsilon=0.25075, gamma=scale, kernel=sigmoid,  
shrinking=False; total time= 0.7s  
[CV] END C=0.1, degree=4, epsilon=0.25075, gamma=scale, kernel=sigmoid,  
shrinking=False; total time= 0.7s  
[CV] END C=0.1, degree=4, epsilon=0.25075, gamma=scale, kernel=sigmoid,  
shrinking=False; total time= 0.7s  
[CV] END C=1, degree=5, epsilon=0.75025, gamma=auto, kernel=poly,  
shrinking=False; total time= 3.4s  
[CV] END C=1, degree=5, epsilon=0.75025, gamma=auto, kernel=poly,  
shrinking=False; total time= 2.7s  
[CV] END C=1, degree=5, epsilon=0.75025, gamma=auto, kernel=poly,  
shrinking=False; total time= 2.9s  
[CV] END C=1, degree=5, epsilon=0.75025, gamma=auto, kernel=poly,  
shrinking=False; total time= 3.4s
```

```
[CV] END C=1, degree=5, epsilon=0.75025, gamma=auto, kernel=poly,  
shrinking=False; total time= 2.9s  
[CV] END C=1, degree=2, epsilon=0.75025, gamma=auto, kernel=poly,  
shrinking=False; total time= 0.4s  
[CV] END C=1, degree=2, epsilon=0.75025, gamma=auto, kernel=poly,  
shrinking=False; total time= 0.3s  
[CV] END C=1, degree=2, epsilon=0.75025, gamma=auto, kernel=poly,  
shrinking=False; total time= 0.3s  
[CV] END C=1, degree=2, epsilon=0.75025, gamma=auto, kernel=poly,  
shrinking=False; total time= 0.4s  
[CV] END C=1, degree=2, epsilon=0.75025, gamma=auto, kernel=poly,  
shrinking=False; total time= 0.3s  
[CV] END C=0.1, degree=3, epsilon=0.001, gamma=scale, kernel=rbf,  
shrinking=True; total time= 0.7s  
[CV] END C=0.1, degree=3, epsilon=0.001, gamma=scale, kernel=rbf,  
shrinking=True; total time= 0.7s  
[CV] END C=0.1, degree=3, epsilon=0.001, gamma=scale, kernel=rbf,  
shrinking=True; total time= 0.7s  
[CV] END C=0.1, degree=3, epsilon=0.001, gamma=scale, kernel=rbf,  
shrinking=True; total time= 0.7s  
[CV] END C=0.1, degree=3, epsilon=0.001, gamma=scale, kernel=rbf,  
shrinking=True; total time= 1.2s  
[CV] END C=0.1, degree=2, epsilon=0.5005, gamma=auto, kernel=rbf,  
shrinking=False; total time= 0.7s  
[CV] END C=0.1, degree=2, epsilon=0.5005, gamma=auto, kernel=rbf,  
shrinking=False; total time= 0.4s  
[CV] END C=0.1, degree=2, epsilon=0.5005, gamma=auto, kernel=rbf,  
shrinking=False; total time= 0.4s  
[CV] END C=0.1, degree=2, epsilon=0.5005, gamma=auto, kernel=rbf,  
shrinking=False; total time= 0.4s  
[CV] END C=0.001, degree=5, epsilon=0.25075, gamma=auto, kernel=linear,  
shrinking=False; total time= 0.2s  
[CV] END C=0.001, degree=5, epsilon=0.25075, gamma=auto, kernel=linear,  
shrinking=False; total time= 0.2s  
[CV] END C=0.001, degree=5, epsilon=0.25075, gamma=auto, kernel=linear,  
shrinking=False; total time= 0.4s  
[CV] END C=0.001, degree=5, epsilon=0.25075, gamma=auto, kernel=linear,  
shrinking=False; total time= 0.7s  
[CV] END C=0.001, degree=5, epsilon=0.25075, gamma=auto, kernel=linear,  
shrinking=False; total time= 0.7s  
[CV] END C=0.01, degree=5, epsilon=0.25075, gamma=auto, kernel=poly,  
shrinking=True; total time= 0.4s  
[CV] END C=0.01, degree=5, epsilon=0.25075, gamma=auto, kernel=poly,  
shrinking=True; total time= 0.3s  
[CV] END C=0.01, degree=5, epsilon=0.25075, gamma=auto, kernel=poly,  
shrinking=True; total time= 0.3s
```

```
[CV] END C=0.01, degree=5, epsilon=0.25075, gamma=auto, kernel=poly,  
shrinking=True; total time= 0.3s  
[CV] END C=0.01, degree=5, epsilon=0.25075, gamma=auto, kernel=poly,  
shrinking=True; total time= 0.3s  
Mean Absolute Percentage Error(MAPE): 2.0159023659059336  
Root Mean Squared Error(RMSE): 0.8061338824665089  
R2 Score: 0.36517067123023694
```





```
[96]: grid_svr.best_params_
```

```
[96]: {'shrinking': True,
       'kernel': 'rbf',
       'gamma': 'scale',
       'epsilon': 0.25075,
       'degree': 3,
       'C': 1}
```

```
[97]: param_grid = {'loss': ['epsilon_insensitive','squared_epsilon_insensitive'],
                  'C': [0.0001,0.001,0.01,0.1,1],
                  'epsilon': np.linspace(0.001,1,5)}

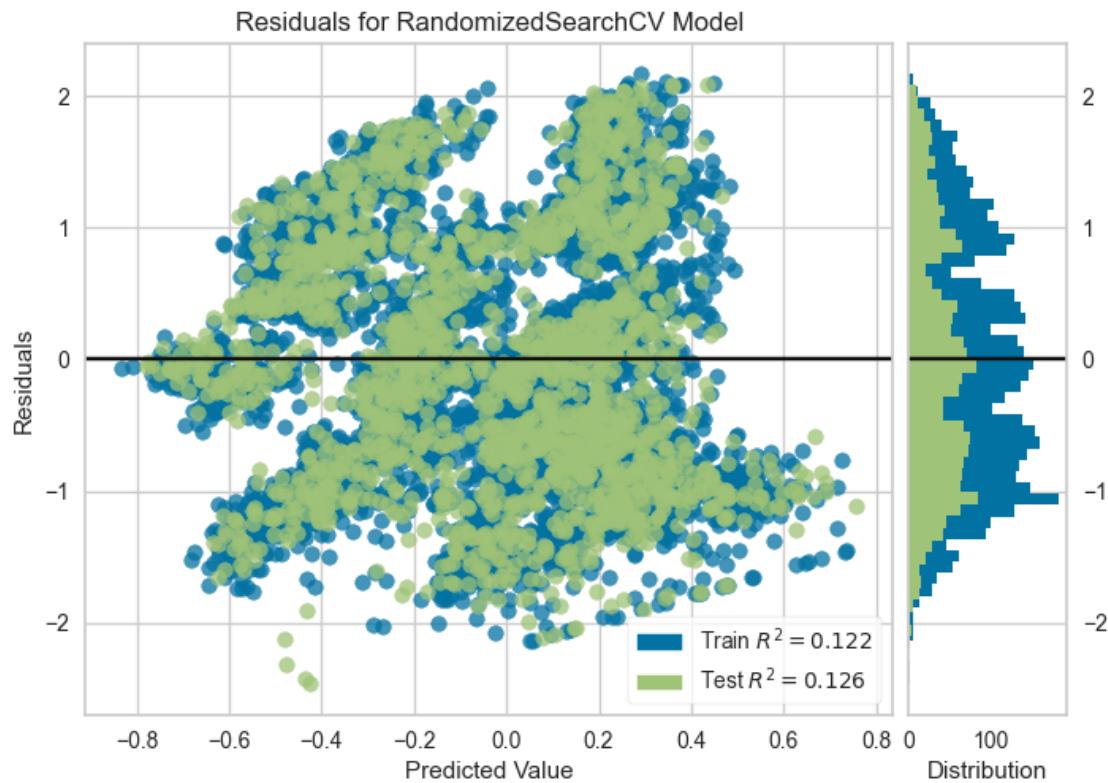
grid_lsvr = RandomizedSearchCV(LinearSVR(),param_grid, cv=RepeatedKFold(n_splits=6,n_repeats=2),verbose=1)
train_and_evaluate_model(grid_lsvr)
```

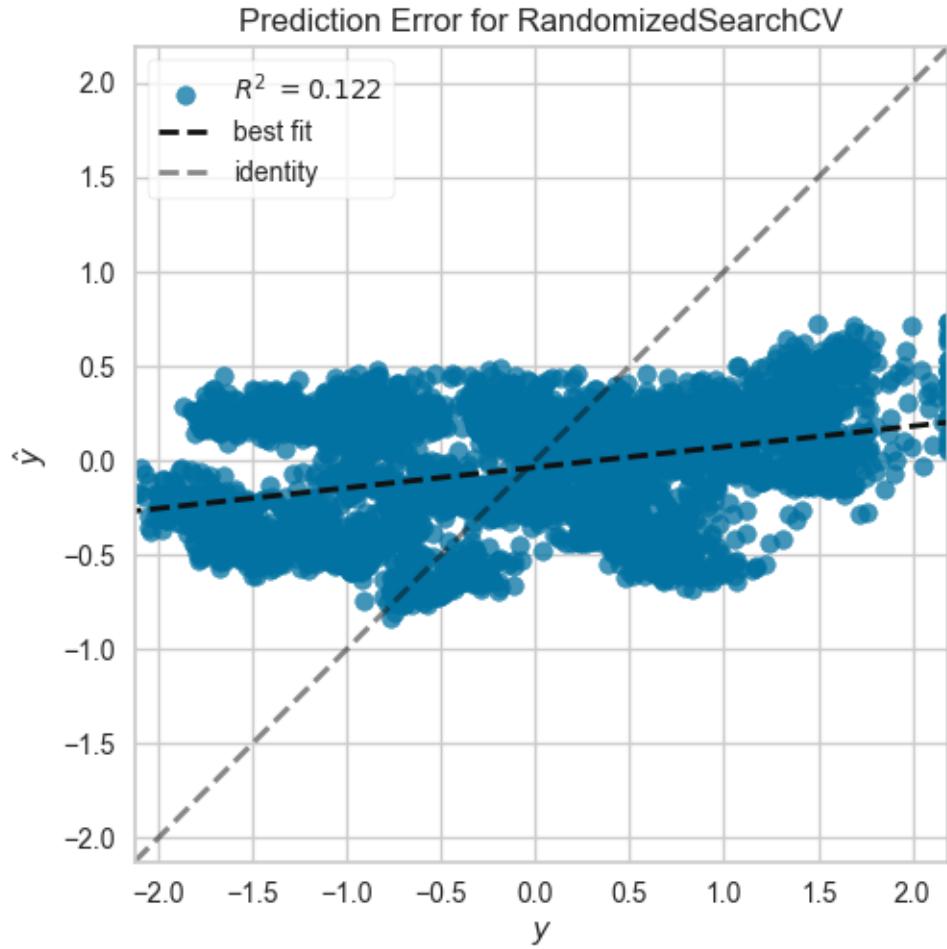






```
0.0s
[CV] END C=0.001, epsilon=0.5005, loss=squared_epsilon_insensitive; total time=
0.0s
[CV] END ...C=1, epsilon=1.0, loss=epsilon_insensitive; total time= 0.0s
[CV] END ...C=1, epsilon=1.0, loss=epsilon_insensitive; total time= 0.0s
[CV] END ...C=1, epsilon=1.0, loss=epsilon_insensitive; total time= 0.0s
[CV] END ...C=1, epsilon=1.0, loss=epsilon_insensitive; total time= 0.0s
[CV] END ...C=1, epsilon=1.0, loss=epsilon_insensitive; total time= 0.0s
[CV] END ...C=1, epsilon=1.0, loss=epsilon_insensitive; total time= 0.0s
[CV] END ...C=1, epsilon=1.0, loss=epsilon_insensitive; total time= 0.0s
[CV] END ...C=1, epsilon=1.0, loss=epsilon_insensitive; total time= 0.0s
[CV] END ...C=1, epsilon=1.0, loss=epsilon_insensitive; total time= 0.0s
[CV] END ...C=1, epsilon=1.0, loss=epsilon_insensitive; total time= 0.0s
[CV] END ...C=1, epsilon=1.0, loss=epsilon_insensitive; total time= 0.0s
[CV] END ...C=1, epsilon=1.0, loss=epsilon_insensitive; total time= 0.0s
[CV] END ...C=1, epsilon=1.0, loss=epsilon_insensitive; total time= 0.0s
[CV] END ...C=1, epsilon=1.0, loss=epsilon_insensitive; total time= 0.0s
Mean Absolute Percentage Error(MAPE): 1.8027668839748314
Root Mean Squared Error(RMSE): 0.9457010405217784
R2 Score: 0.12632408018212493
```





```
[98]: grid_lsvr.best_params_
```

```
[98]: {'loss': 'squared_epsilon_insensitive', 'epsilon': 0.5005, 'C': 0.1}
```

```
[99]: param_grid = {'criterion': [
    'squared_error', 'friedman_mse', 'absolute_error', 'poisson'],
    'splitter': ['best', 'random'],
    'max_features': ['auto', 'sqrt', 'log2']
}
```

```
grid_dt = RandomizedSearchCV(DecisionTreeRegressor(), param_grid, verbose=2, cv=RepeatedKFold(n_splits=6))
train_and_evaluate_model(grid_dt)
```

Fitting 18 folds for each of 10 candidates, totalling 180 fits

[CV] END criterion=poisson, max\_features=log2, splitter=best; total time= 0.0s

[CV] END criterion=poisson, max\_features=log2, splitter=best; total time= 0.0s





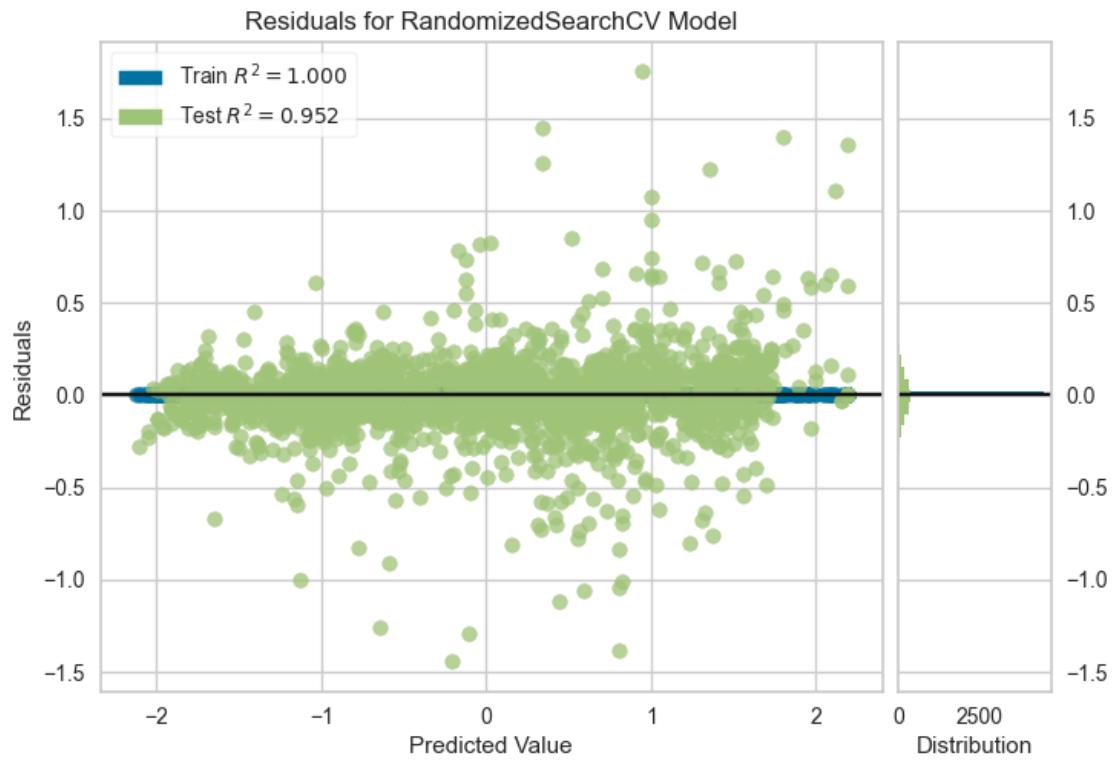


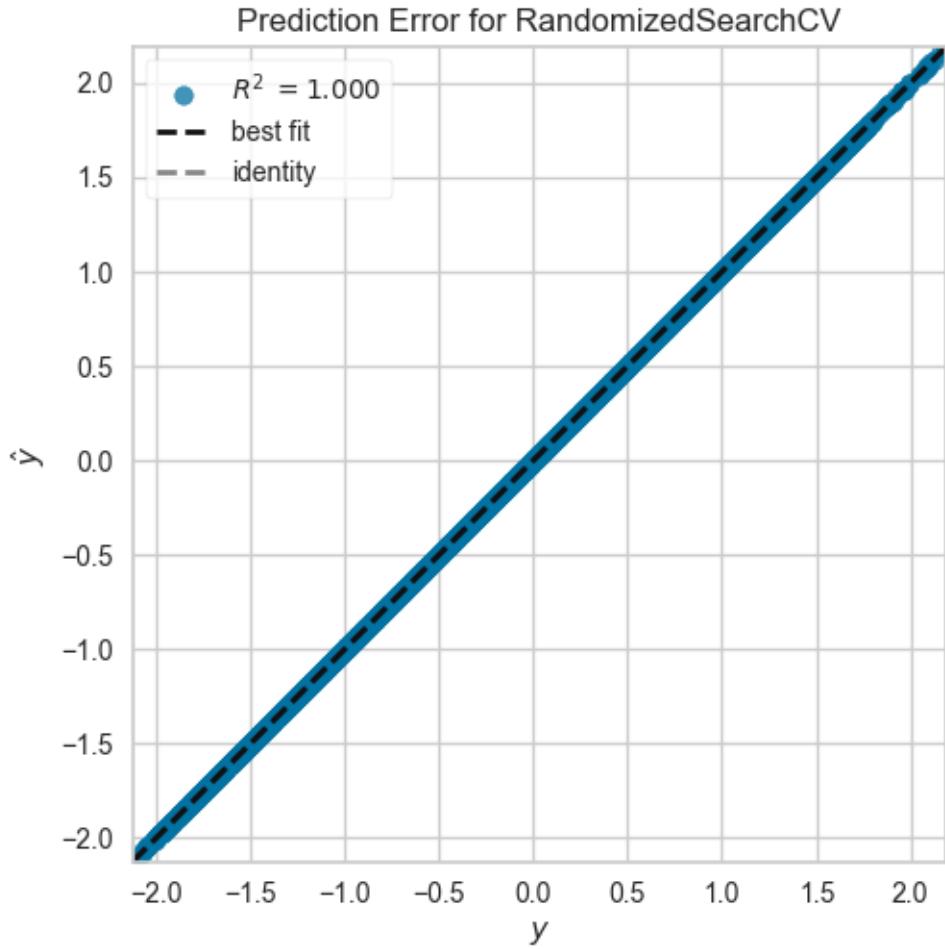






```
[CV] END criterion=absolute_error, max_features=sqrt, splitter=best; total time= 0.0s
[CV] END criterion=absolute_error, max_features=sqrt, splitter=best; total time= 0.1s
[CV] END criterion=absolute_error, max_features=sqrt, splitter=best; total time= 0.0s
[CV] END criterion=absolute_error, max_features=sqrt, splitter=best; total time= 0.1s
[CV] END criterion=absolute_error, max_features=sqrt, splitter=best; total time= 0.0s
[CV] END criterion=absolute_error, max_features=sqrt, splitter=best; total time= 0.1s
[CV] END criterion=absolute_error, max_features=sqrt, splitter=best; total time= 0.0s
Mean Absolute Percentage Error(MAPE): 0.6790385905167696
Root Mean Squared Error(RMSE): 0.22065466321116575
R2 Score: 0.9524370424226405
```





```
[100]: grid_dt.best_params_
```

```
[100]: {'splitter': 'best', 'max_features': 'auto', 'criterion': 'friedman_mse'}
```

```
[101]: param_grid = {'loss': ['squared_error','absolute_error','huber','quantile'],
                  'n_estimators': [100,400,700,1000],
                  'learning_rate': [0.2,0.4,0.7,1],
                  'criterion': ['friedman_mse','squared_error'],
                  'max_features': ['auto','sqrt','log2']}
      }
```

```
grid_gb = ↵
    ↵RandomizedSearchCV(GradientBoostingRegressor(),param_grid,verbose=3,cv=6)
train_and_evaluate_model(grid_gb)
```

Fitting 6 folds for each of 10 candidates, totalling 60 fits  
[CV 1/6] END criterion=friedman\_mse, learning\_rate=0.4, loss=absolute\_error,

```

max_features=log2, n_estimators=1000;, score=0.969 total time= 3.5s
[CV 2/6] END criterion=friedman_mse, learning_rate=0.4, loss=absolute_error,
max_features=log2, n_estimators=1000;, score=0.976 total time= 3.4s
[CV 3/6] END criterion=friedman_mse, learning_rate=0.4, loss=absolute_error,
max_features=log2, n_estimators=1000;, score=0.960 total time= 3.6s
[CV 4/6] END criterion=friedman_mse, learning_rate=0.4, loss=absolute_error,
max_features=log2, n_estimators=1000;, score=0.962 total time= 3.6s
[CV 5/6] END criterion=friedman_mse, learning_rate=0.4, loss=absolute_error,
max_features=log2, n_estimators=1000;, score=0.967 total time= 3.5s
[CV 6/6] END criterion=friedman_mse, learning_rate=0.4, loss=absolute_error,
max_features=log2, n_estimators=1000;, score=0.960 total time= 3.6s
[CV 1/6] END criterion=squared_error, learning_rate=0.2, loss=quantile,
max_features=log2, n_estimators=400;, score=0.582 total time= 1.6s
[CV 2/6] END criterion=squared_error, learning_rate=0.2, loss=quantile,
max_features=log2, n_estimators=400;, score=0.514 total time= 1.7s
[CV 3/6] END criterion=squared_error, learning_rate=0.2, loss=quantile,
max_features=log2, n_estimators=400;, score=0.495 total time= 1.7s
[CV 4/6] END criterion=squared_error, learning_rate=0.2, loss=quantile,
max_features=log2, n_estimators=400;, score=0.628 total time= 1.6s
[CV 5/6] END criterion=squared_error, learning_rate=0.2, loss=quantile,
max_features=log2, n_estimators=400;, score=0.479 total time= 1.6s
[CV 6/6] END criterion=squared_error, learning_rate=0.2, loss=quantile,
max_features=log2, n_estimators=400;, score=0.543 total time= 1.4s
[CV 1/6] END criterion=friedman_mse, learning_rate=0.2, loss=squared_error,
max_features=auto, n_estimators=100;, score=0.966 total time= 2.2min
[CV 2/6] END criterion=friedman_mse, learning_rate=0.2, loss=squared_error,
max_features=auto, n_estimators=100;, score=0.972 total time= 0.5s
[CV 3/6] END criterion=friedman_mse, learning_rate=0.2, loss=squared_error,
max_features=auto, n_estimators=100;, score=0.964 total time= 0.5s
[CV 4/6] END criterion=friedman_mse, learning_rate=0.2, loss=squared_error,
max_features=auto, n_estimators=100;, score=0.962 total time= 0.7s
[CV 5/6] END criterion=friedman_mse, learning_rate=0.2, loss=squared_error,
max_features=auto, n_estimators=100;, score=0.970 total time= 1.4s
[CV 6/6] END criterion=friedman_mse, learning_rate=0.2, loss=squared_error,
max_features=auto, n_estimators=100;, score=0.968 total time= 0.8s
[CV 1/6] END criterion=friedman_mse, learning_rate=0.2, loss=quantile,
max_features=auto, n_estimators=700;, score=0.934 total time= 6.4s
[CV 2/6] END criterion=friedman_mse, learning_rate=0.2, loss=quantile,
max_features=auto, n_estimators=700;, score=0.890 total time= 4.7s
[CV 3/6] END criterion=friedman_mse, learning_rate=0.2, loss=quantile,
max_features=auto, n_estimators=700;, score=0.881 total time= 4.6s
[CV 4/6] END criterion=friedman_mse, learning_rate=0.2, loss=quantile,
max_features=auto, n_estimators=700;, score=0.907 total time= 4.8s
[CV 5/6] END criterion=friedman_mse, learning_rate=0.2, loss=quantile,
max_features=auto, n_estimators=700;, score=0.922 total time= 4.6s
[CV 6/6] END criterion=friedman_mse, learning_rate=0.2, loss=quantile,
max_features=auto, n_estimators=700;, score=0.884 total time= 4.7s
[CV 1/6] END criterion=squared_error, learning_rate=1, loss=squared_error,

```

```

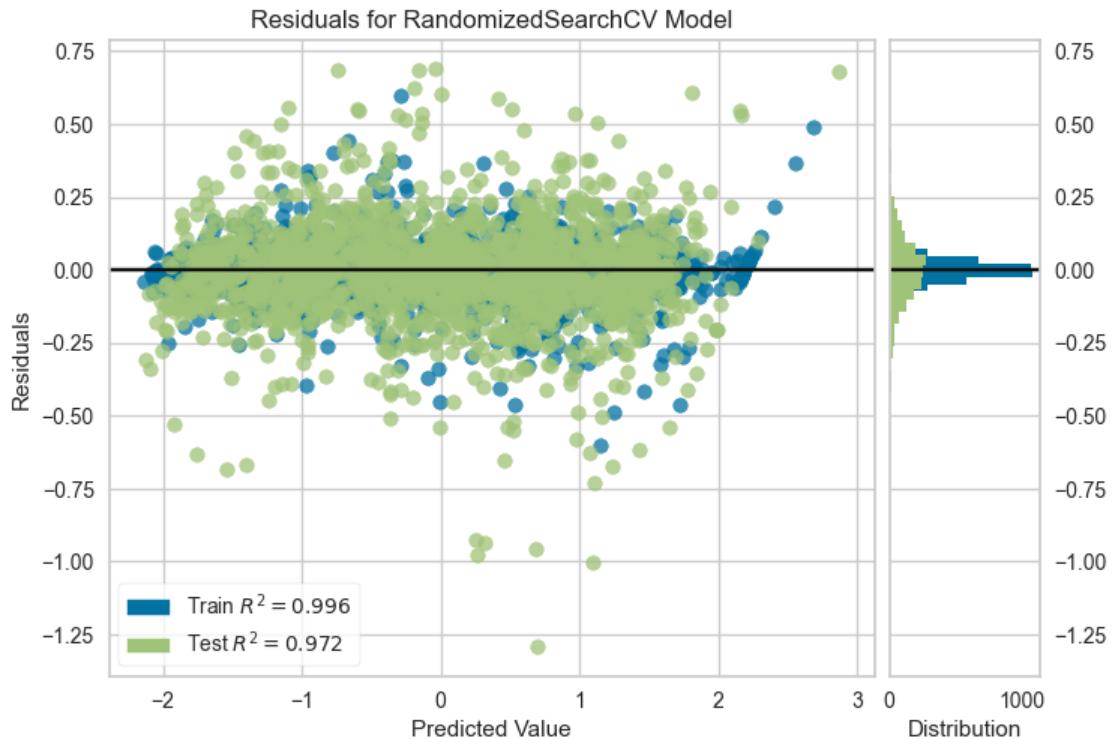
max_features=sqrt, n_estimators=100;, score=0.932 total time= 0.1s
[CV 2/6] END criterion=squared_error, learning_rate=1, loss=squared_error,
max_features=sqrt, n_estimators=100;, score=0.943 total time= 0.1s
[CV 3/6] END criterion=squared_error, learning_rate=1, loss=squared_error,
max_features=sqrt, n_estimators=100;, score=0.913 total time= 0.1s
[CV 4/6] END criterion=squared_error, learning_rate=1, loss=squared_error,
max_features=sqrt, n_estimators=100;, score=0.899 total time= 0.1s
[CV 5/6] END criterion=squared_error, learning_rate=1, loss=squared_error,
max_features=sqrt, n_estimators=100;, score=0.943 total time= 0.1s
[CV 6/6] END criterion=squared_error, learning_rate=1, loss=squared_error,
max_features=sqrt, n_estimators=100;, score=0.882 total time= 0.1s
[CV 1/6] END criterion=squared_error, learning_rate=0.7, loss=huber,
max_features=sqrt, n_estimators=1000;, score=0.969 total time= 4.4s
[CV 2/6] END criterion=squared_error, learning_rate=0.7, loss=huber,
max_features=sqrt, n_estimators=1000;, score=0.964 total time= 4.4s
[CV 3/6] END criterion=squared_error, learning_rate=0.7, loss=huber,
max_features=sqrt, n_estimators=1000;, score=0.970 total time= 4.4s
[CV 4/6] END criterion=squared_error, learning_rate=0.7, loss=huber,
max_features=sqrt, n_estimators=1000;, score=0.962 total time= 4.7s
[CV 5/6] END criterion=squared_error, learning_rate=0.7, loss=huber,
max_features=sqrt, n_estimators=1000;, score=0.975 total time= 4.4s
[CV 6/6] END criterion=squared_error, learning_rate=0.7, loss=huber,
max_features=sqrt, n_estimators=1000;, score=0.962 total time= 4.4s
[CV 1/6] END criterion=squared_error, learning_rate=0.7, loss=squared_error,
max_features=log2, n_estimators=700;, score=0.967 total time= 1.2s
[CV 2/6] END criterion=squared_error, learning_rate=0.7, loss=squared_error,
max_features=log2, n_estimators=700;, score=0.969 total time= 1.2s
[CV 3/6] END criterion=squared_error, learning_rate=0.7, loss=squared_error,
max_features=log2, n_estimators=700;, score=0.960 total time= 1.6s
[CV 4/6] END criterion=squared_error, learning_rate=0.7, loss=squared_error,
max_features=log2, n_estimators=700;, score=0.961 total time= 1.5s
[CV 5/6] END criterion=squared_error, learning_rate=0.7, loss=squared_error,
max_features=log2, n_estimators=700;, score=0.971 total time= 1.5s
[CV 6/6] END criterion=squared_error, learning_rate=0.7, loss=squared_error,
max_features=log2, n_estimators=700;, score=0.969 total time= 1.4s
[CV 1/6] END criterion=friedman_mse, learning_rate=0.7, loss=huber,
max_features=sqrt, n_estimators=400;, score=0.967 total time= 1.7s
[CV 2/6] END criterion=friedman_mse, learning_rate=0.7, loss=huber,
max_features=sqrt, n_estimators=400;, score=0.959 total time= 1.7s
[CV 3/6] END criterion=friedman_mse, learning_rate=0.7, loss=huber,
max_features=sqrt, n_estimators=400;, score=0.953 total time= 1.7s
[CV 4/6] END criterion=friedman_mse, learning_rate=0.7, loss=huber,
max_features=sqrt, n_estimators=400;, score=0.963 total time= 1.7s
[CV 5/6] END criterion=friedman_mse, learning_rate=0.7, loss=huber,
max_features=sqrt, n_estimators=400;, score=0.966 total time= 1.7s
[CV 6/6] END criterion=friedman_mse, learning_rate=0.7, loss=huber,
max_features=sqrt, n_estimators=400;, score=0.964 total time= 1.7s
[CV 1/6] END criterion=squared_error, learning_rate=0.2, loss=absolute_error,

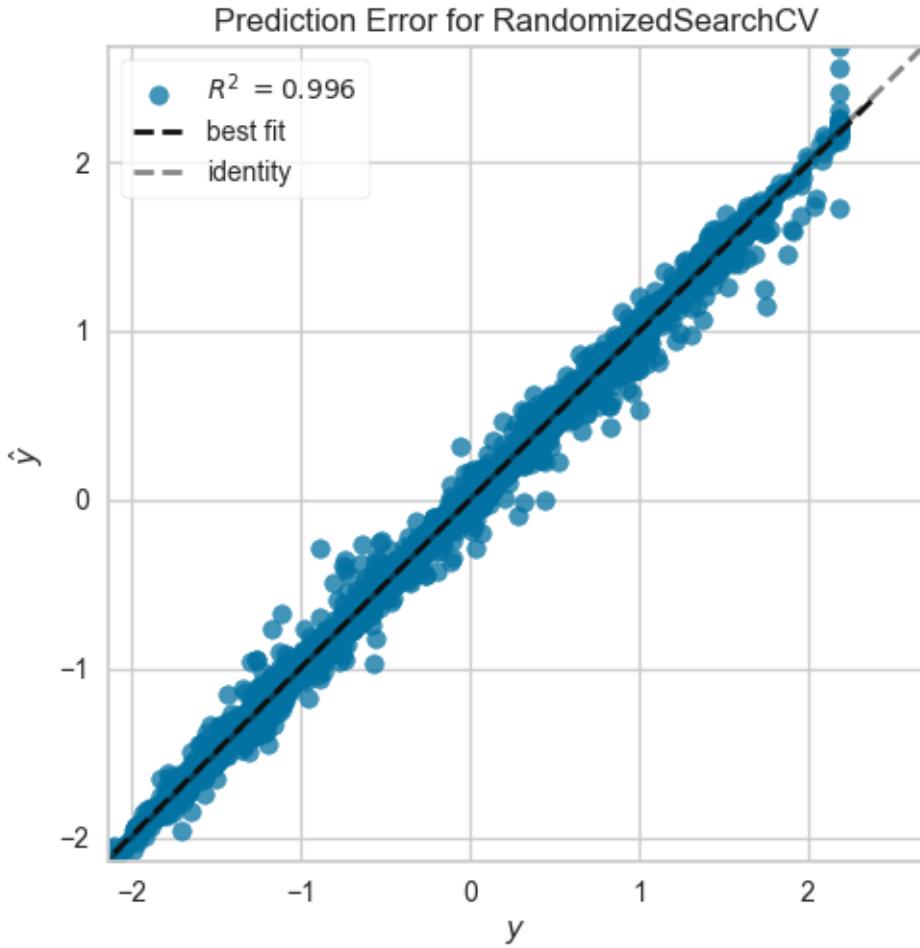
```

```

max_features=log2, n_estimators=700;, score=0.960 total time= 2.5s
[CV 2/6] END criterion=squared_error, learning_rate=0.2, loss=absolute_error,
max_features=log2, n_estimators=700;, score=0.966 total time= 2.5s
[CV 3/6] END criterion=squared_error, learning_rate=0.2, loss=absolute_error,
max_features=log2, n_estimators=700;, score=0.961 total time= 2.5s
[CV 4/6] END criterion=squared_error, learning_rate=0.2, loss=absolute_error,
max_features=log2, n_estimators=700;, score=0.949 total time= 2.5s
[CV 5/6] END criterion=squared_error, learning_rate=0.2, loss=absolute_error,
max_features=log2, n_estimators=700;, score=0.966 total time= 2.6s
[CV 6/6] END criterion=squared_error, learning_rate=0.2, loss=absolute_error,
max_features=log2, n_estimators=700;, score=0.964 total time= 2.5s
[CV 1/6] END criterion=squared_error, learning_rate=1, loss=huber,
max_features=log2, n_estimators=100;, score=0.931 total time= 0.4s
[CV 2/6] END criterion=squared_error, learning_rate=1, loss=huber,
max_features=log2, n_estimators=100;, score=0.894 total time= 0.4s
[CV 3/6] END criterion=squared_error, learning_rate=1, loss=huber,
max_features=log2, n_estimators=100;, score=0.935 total time= 0.4s
[CV 4/6] END criterion=squared_error, learning_rate=1, loss=huber,
max_features=log2, n_estimators=100;, score=0.927 total time= 0.4s
[CV 5/6] END criterion=squared_error, learning_rate=1, loss=huber,
max_features=log2, n_estimators=100;, score=0.922 total time= 0.4s
[CV 6/6] END criterion=squared_error, learning_rate=1, loss=huber,
max_features=log2, n_estimators=100;, score=0.951 total time= 0.4s
Mean Absolute Percentage Error(MAPE): 0.5385070919758704
Root Mean Squared Error(RMSE): 0.16830518927732563
R2 Score: 0.9723281850145602

```





```
[102]: grid_gb.best_params_
```

```
[102]: {'n_estimators': 1000,
      'max_features': 'sqrt',
      'loss': 'huber',
      'learning_rate': 0.7,
      'criterion': 'squared_error'}
```

```
[103]: param_grid = {'learning_rate': [0.2, 0.4, 0.5, 0.7, 1],
                  'n_estimators': [100, 400, 700, 800, 1000]
                 }
```

```
grid_cat = RandomizedSearchCV(CatBoostRegressor(silent=True), param_grid, verbose=5, cv=5)
```

```
train_and_evaluate_model(grid_cat)
```

```
Fitting 5 folds for each of 10 candidates, totalling 50 fits
[CV 1/5] END learning_rate=0.4, n_estimators=700;, score=0.986 total time=
1.7s
[CV 2/5] END learning_rate=0.4, n_estimators=700;, score=0.986 total time=
2.1s
[CV 3/5] END learning_rate=0.4, n_estimators=700;, score=0.984 total time=
1.6s
[CV 4/5] END learning_rate=0.4, n_estimators=700;, score=0.986 total time=
1.6s
[CV 5/5] END learning_rate=0.4, n_estimators=700;, score=0.986 total time=
1.9s
[CV 1/5] END learning_rate=0.2, n_estimators=800;, score=0.987 total time=
1.9s
[CV 2/5] END learning_rate=0.2, n_estimators=800;, score=0.986 total time=
1.8s
[CV 3/5] END learning_rate=0.2, n_estimators=800;, score=0.984 total time=
2.0s
[CV 4/5] END learning_rate=0.2, n_estimators=800;, score=0.986 total time=
1.9s
[CV 5/5] END learning_rate=0.2, n_estimators=800;, score=0.987 total time=
2.1s
[CV 1/5] END .learning_rate=1, n_estimators=700;, score=0.978 total time= 2.1s
[CV 2/5] END .learning_rate=1, n_estimators=700;, score=0.980 total time= 1.9s
[CV 3/5] END .learning_rate=1, n_estimators=700;, score=0.974 total time= 2.2s
[CV 4/5] END .learning_rate=1, n_estimators=700;, score=0.972 total time= 2.3s
[CV 5/5] END .learning_rate=1, n_estimators=700;, score=0.977 total time= 2.9s
[CV 1/5] END learning_rate=1, n_estimators=1000;, score=0.978 total time= 4.4s
[CV 2/5] END learning_rate=1, n_estimators=1000;, score=0.980 total time= 4.2s
[CV 3/5] END learning_rate=1, n_estimators=1000;, score=0.974 total time= 2.5s
[CV 4/5] END learning_rate=1, n_estimators=1000;, score=0.972 total time= 2.4s
[CV 5/5] END learning_rate=1, n_estimators=1000;, score=0.977 total time= 2.3s
[CV 1/5] END learning_rate=0.2, n_estimators=400;, score=0.986 total time=
1.5s
[CV 2/5] END learning_rate=0.2, n_estimators=400;, score=0.985 total time=
2.4s
[CV 3/5] END learning_rate=0.2, n_estimators=400;, score=0.983 total time=
2.3s
[CV 4/5] END learning_rate=0.2, n_estimators=400;, score=0.985 total time=
2.3s
[CV 5/5] END learning_rate=0.2, n_estimators=400;, score=0.986 total time=
2.3s
[CV 1/5] END learning_rate=0.2, n_estimators=700;, score=0.987 total time=
2.6s
[CV 2/5] END learning_rate=0.2, n_estimators=700;, score=0.986 total time=
2.2s
[CV 3/5] END learning_rate=0.2, n_estimators=700;, score=0.984 total time=
```

```

2.0s
[CV 4/5] END learning_rate=0.2, n_estimators=700;, score=0.986 total time=
2.0s
[CV 5/5] END learning_rate=0.2, n_estimators=700;, score=0.987 total time=
2.1s
[CV 1/5] END learning_rate=0.7, n_estimators=700;, score=0.983 total time=
2.4s
[CV 2/5] END learning_rate=0.7, n_estimators=700;, score=0.983 total time=
2.4s
[CV 3/5] END learning_rate=0.7, n_estimators=700;, score=0.978 total time=
2.2s
[CV 4/5] END learning_rate=0.7, n_estimators=700;, score=0.983 total time=
1.8s
[CV 5/5] END learning_rate=0.7, n_estimators=700;, score=0.981 total time=
1.8s
[CV 1/5] END learning_rate=0.7, n_estimators=400;, score=0.983 total time=
1.7s
[CV 2/5] END learning_rate=0.7, n_estimators=400;, score=0.983 total time=
1.6s
[CV 3/5] END learning_rate=0.7, n_estimators=400;, score=0.978 total time=
1.4s
[CV 4/5] END learning_rate=0.7, n_estimators=400;, score=0.983 total time=
2.2s
[CV 5/5] END learning_rate=0.7, n_estimators=400;, score=0.982 total time=
1.5s
[CV 1/5] END .learning_rate=1, n_estimators=100;, score=0.976 total time= 0.9s
[CV 2/5] END .learning_rate=1, n_estimators=100;, score=0.980 total time= 0.9s
[CV 3/5] END .learning_rate=1, n_estimators=100;, score=0.973 total time= 1.1s
[CV 4/5] END .learning_rate=1, n_estimators=100;, score=0.972 total time= 1.2s
[CV 5/5] END .learning_rate=1, n_estimators=100;, score=0.977 total time= 1.1s
[CV 1/5] END learning_rate=0.4, n_estimators=100;, score=0.980 total time=
1.0s
[CV 2/5] END learning_rate=0.4, n_estimators=100;, score=0.982 total time=
0.9s
[CV 3/5] END learning_rate=0.4, n_estimators=100;, score=0.978 total time=
1.2s
[CV 4/5] END learning_rate=0.4, n_estimators=100;, score=0.980 total time=
1.0s
[CV 5/5] END learning_rate=0.4, n_estimators=100;, score=0.981 total time=
0.9s
Mean Absolute Percentage Error(MAPE): 0.4397850247521942
Root Mean Squared Error(RMSE): 0.11688404578618586
R2 Score: 0.9866539390601259

```

[104]: grid\_cat.best\_params\_

[104]: {'n\_estimators': 800, 'learning\_rate': 0.2}

```
[105]: param_grid = {'n_estimators': [100,300,500,800,1000],
                   'max_samples': [0.24,0.58,0.71,0.96],
                   'max_features': np.linspace(0,1,5),
                   'bootstrap': [True,False],
                   'oob_score': [True,False]
                  }
grid_bag = RandomizedSearchCV(BaggingRegressor(),param_grid,verbose=2,cv=5)
train_and_evaluate_model(grid_bag)
```

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

[CV] END bootstrap=True, max\_features=0.25, max\_samples=0.24, n\_estimators=300, oob\_score=True; total time= 0.9s

[CV] END bootstrap=True, max\_features=0.25, max\_samples=0.24, n\_estimators=300, oob\_score=True; total time= 0.8s

[CV] END bootstrap=True, max\_features=0.25, max\_samples=0.24, n\_estimators=300, oob\_score=True; total time= 0.8s

[CV] END bootstrap=True, max\_features=0.25, max\_samples=0.24, n\_estimators=300, oob\_score=True; total time= 0.8s

[CV] END bootstrap=True, max\_features=0.25, max\_samples=0.24, n\_estimators=300, oob\_score=True; total time= 0.8s

[CV] END bootstrap=False, max\_features=0.5, max\_samples=0.24, n\_estimators=800, oob\_score=True; total time= 0.0s

[CV] END bootstrap=False, max\_features=0.5, max\_samples=0.24, n\_estimators=800, oob\_score=True; total time= 0.0s

[CV] END bootstrap=False, max\_features=0.5, max\_samples=0.24, n\_estimators=800, oob\_score=True; total time= 0.0s

[CV] END bootstrap=False, max\_features=0.5, max\_samples=0.24, n\_estimators=800, oob\_score=True; total time= 0.0s

[CV] END bootstrap=False, max\_features=0.75, max\_samples=0.96, n\_estimators=500, oob\_score=False; total time= 10.5s

[CV] END bootstrap=False, max\_features=0.75, max\_samples=0.96, n\_estimators=500, oob\_score=False; total time= 10.4s

[CV] END bootstrap=False, max\_features=0.75, max\_samples=0.96, n\_estimators=500, oob\_score=False; total time= 10.5s

[CV] END bootstrap=False, max\_features=0.75, max\_samples=0.96, n\_estimators=500, oob\_score=False; total time= 10.4s

[CV] END bootstrap=False, max\_features=0.75, max\_samples=0.96, n\_estimators=500, oob\_score=False; total time= 11.5s

[CV] END bootstrap=True, max\_features=1.0, max\_samples=0.24, n\_estimators=500, oob\_score=True; total time= 3.7s

[CV] END bootstrap=True, max\_features=1.0, max\_samples=0.24, n\_estimators=500, oob\_score=True; total time= 3.6s

[CV] END bootstrap=True, max\_features=1.0, max\_samples=0.24, n\_estimators=500, oob\_score=True; total time= 3.6s

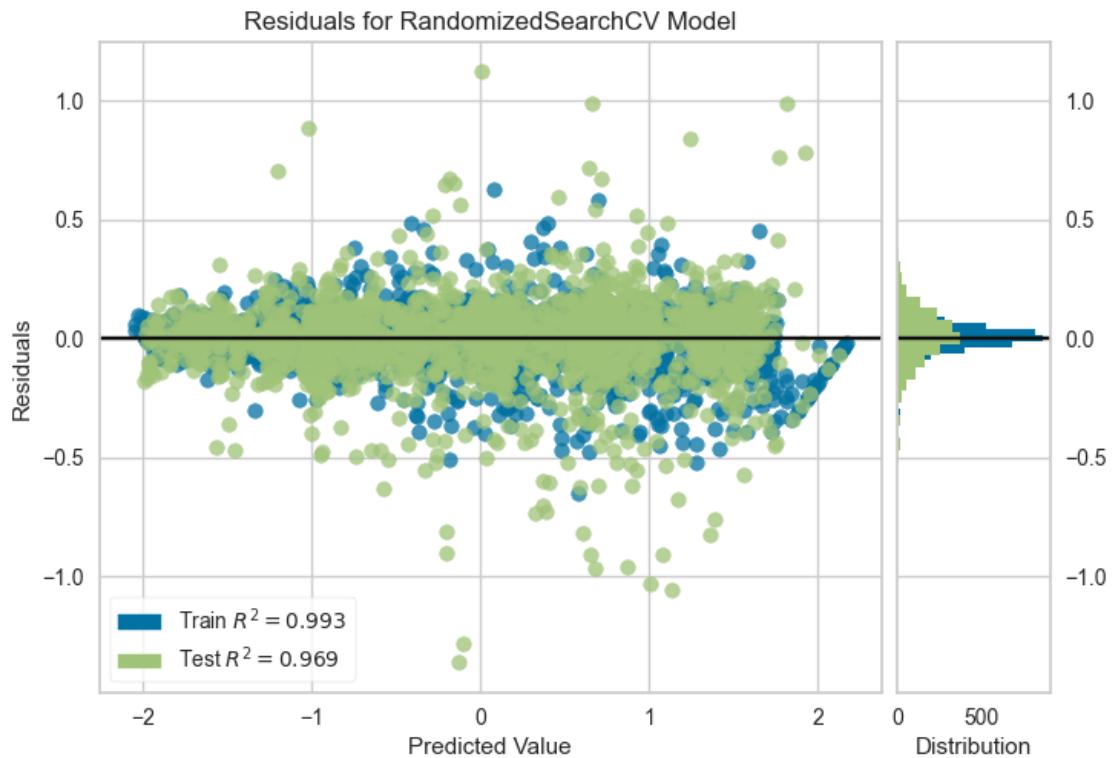
[CV] END bootstrap=True, max\_features=1.0, max\_samples=0.24, n\_estimators=500, oob\_score=True; total time= 3.6s

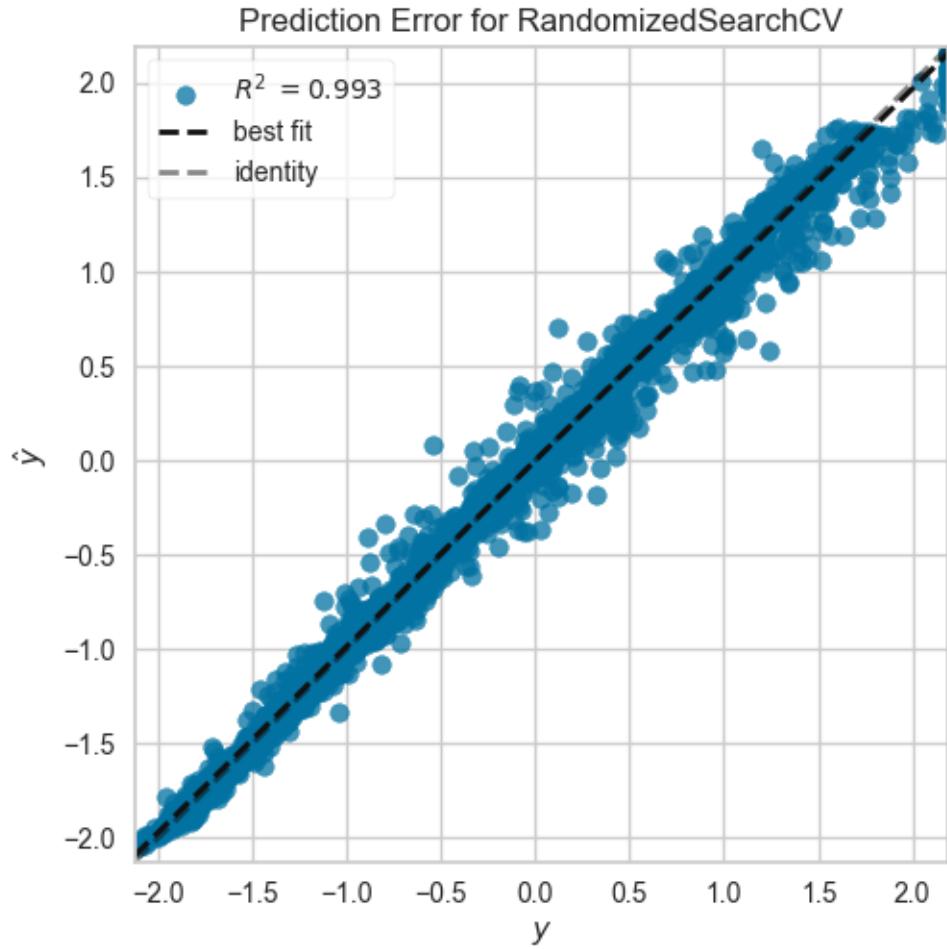
```
[CV] END bootstrap=True, max_features=1.0, max_samples=0.24, n_estimators=500,
oob_score=True; total time= 3.6s
[CV] END bootstrap=False, max_features=0.25, max_samples=0.71, n_estimators=500,
oob_score=False; total time= 2.2s
[CV] END bootstrap=False, max_features=0.25, max_samples=0.71, n_estimators=500,
oob_score=False; total time= 2.1s
[CV] END bootstrap=False, max_features=0.25, max_samples=0.71, n_estimators=500,
oob_score=False; total time= 2.2s
[CV] END bootstrap=False, max_features=0.25, max_samples=0.71, n_estimators=500,
oob_score=False; total time= 2.0s
[CV] END bootstrap=False, max_features=0.25, max_samples=0.71, n_estimators=500,
oob_score=False; total time= 1.9s
[CV] END bootstrap=False, max_features=0.75, max_samples=0.58, n_estimators=800,
oob_score=False; total time= 10.8s
[CV] END bootstrap=False, max_features=0.75, max_samples=0.58, n_estimators=800,
oob_score=False; total time= 10.5s
[CV] END bootstrap=False, max_features=0.75, max_samples=0.58, n_estimators=800,
oob_score=False; total time= 10.9s
[CV] END bootstrap=False, max_features=0.75, max_samples=0.58, n_estimators=800,
oob_score=False; total time= 11.2s
[CV] END bootstrap=False, max_features=0.75, max_samples=0.58, n_estimators=800,
oob_score=False; total time= 10.2s
[CV] END bootstrap=False, max_features=0.0, max_samples=0.24, n_estimators=100,
oob_score=True; total time= 0.0s
[CV] END bootstrap=False, max_features=0.0, max_samples=0.24, n_estimators=100,
oob_score=True; total time= 0.0s
[CV] END bootstrap=False, max_features=0.0, max_samples=0.24, n_estimators=100,
oob_score=True; total time= 0.0s
[CV] END bootstrap=False, max_features=0.0, max_samples=0.24, n_estimators=100,
oob_score=True; total time= 0.0s
[CV] END bootstrap=False, max_features=0.0, max_samples=0.24, n_estimators=100,
oob_score=True; total time= 0.0s
[CV] END bootstrap=False, max_features=0.5, max_samples=0.24, n_estimators=800,
oob_score=False; total time= 3.6s
[CV] END bootstrap=False, max_features=0.5, max_samples=0.24, n_estimators=800,
oob_score=False; total time= 3.5s
[CV] END bootstrap=False, max_features=0.5, max_samples=0.24, n_estimators=800,
oob_score=False; total time= 3.6s
[CV] END bootstrap=False, max_features=0.5, max_samples=0.24, n_estimators=800,
oob_score=False; total time= 3.6s
[CV] END bootstrap=False, max_features=1.0, max_samples=0.71, n_estimators=100,
oob_score=False; total time= 1.9s
[CV] END bootstrap=False, max_features=1.0, max_samples=0.71, n_estimators=100,
oob_score=False; total time= 1.9s
[CV] END bootstrap=False, max_features=1.0, max_samples=0.71, n_estimators=100,
oob_score=False; total time= 1.9s
```

```

[CV] END bootstrap=False, max_features=1.0, max_samples=0.71, n_estimators=100,
oob_score=False; total time= 1.9s
[CV] END bootstrap=False, max_features=1.0, max_samples=0.71, n_estimators=100,
oob_score=False; total time= 1.9s
[CV] END bootstrap=True, max_features=1.0, max_samples=0.71, n_estimators=1000,
oob_score=False; total time= 14.1s
[CV] END bootstrap=True, max_features=1.0, max_samples=0.71, n_estimators=1000,
oob_score=False; total time= 15.1s
[CV] END bootstrap=True, max_features=1.0, max_samples=0.71, n_estimators=1000,
oob_score=False; total time= 14.2s
[CV] END bootstrap=True, max_features=1.0, max_samples=0.71, n_estimators=1000,
oob_score=False; total time= 14.2s
[CV] END bootstrap=True, max_features=1.0, max_samples=0.71, n_estimators=1000,
oob_score=False; total time= 14.1s
Mean Absolute Percentage Error(MAPE): 0.6257275100955686
Root Mean Squared Error(RMSE): 0.17932731459094725
R2 Score: 0.9685851115331399

```





```
[106]: grid_bag.best_params_
```

```
[106]: {'oob_score': False,
      'n_estimators': 1000,
      'max_samples': 0.71,
      'max_features': 1.0,
      'bootstrap': True}
```

```
[107]: param_grid = {'boosting_type': ['gbdt', 'dart', 'goss', 'rf'],
                  'learning_rate': np.linspace(0, 1, 6)[1:],
                  'n_estimators': [100, 300, 500, 800, 1000],
                  'importance_type': ['split', 'gain'],
                  'min_split_gain': [0.68, 0.79, 0.87, 1]}
```

```
grid_lgbm = RandomizedSearchCV(LGBMRegressor(), param_grid, verbose=3)
train_and_evaluate_model(grid_lgbm)
```

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

```

[CV 1/5] END boosting_type=goss, importance_type=gain, learning_rate=0.2,
min_split_gain=1, n_estimators=1000;, score=0.967 total time= 0.0s
[CV 2/5] END boosting_type=goss, importance_type=gain, learning_rate=0.2,
min_split_gain=1, n_estimators=1000;, score=0.970 total time= 0.0s
[CV 3/5] END boosting_type=goss, importance_type=gain, learning_rate=0.2,
min_split_gain=1, n_estimators=1000;, score=0.960 total time= 0.0s
[CV 4/5] END boosting_type=goss, importance_type=gain, learning_rate=0.2,
min_split_gain=1, n_estimators=1000;, score=0.966 total time= 0.0s
[CV 5/5] END boosting_type=goss, importance_type=gain, learning_rate=0.2,
min_split_gain=1, n_estimators=1000;, score=0.973 total time= 0.0s
[CV 1/5] END boosting_type=goss, importance_type=split, learning_rate=0.4,
min_split_gain=0.68, n_estimators=500;, score=0.971 total time= 0.0s
[CV 2/5] END boosting_type=goss, importance_type=split, learning_rate=0.4,
min_split_gain=0.68, n_estimators=500;, score=0.970 total time= 0.0s
[CV 3/5] END boosting_type=goss, importance_type=split, learning_rate=0.4,
min_split_gain=0.68, n_estimators=500;, score=0.967 total time= 0.0s
[CV 4/5] END boosting_type=goss, importance_type=split, learning_rate=0.4,
min_split_gain=0.68, n_estimators=500;, score=0.969 total time= 0.0s
[CV 5/5] END boosting_type=goss, importance_type=split, learning_rate=0.4,
min_split_gain=0.68, n_estimators=500;, score=0.971 total time= 0.0s
[CV 1/5] END boosting_type=dart, importance_type=split, learning_rate=1.0,
min_split_gain=1, n_estimators=800;, score=0.951 total time= 0.5s
[CV 2/5] END boosting_type=dart, importance_type=split, learning_rate=1.0,
min_split_gain=1, n_estimators=800;, score=0.941 total time= 0.9s
[CV 3/5] END boosting_type=dart, importance_type=split, learning_rate=1.0,
min_split_gain=1, n_estimators=800;, score=0.935 total time= 0.9s
[CV 4/5] END boosting_type=dart, importance_type=split, learning_rate=1.0,
min_split_gain=1, n_estimators=800;, score=0.940 total time= 0.5s
[CV 5/5] END boosting_type=dart, importance_type=split, learning_rate=1.0,
min_split_gain=1, n_estimators=800;, score=0.963 total time= 0.5s
[CV 1/5] END boosting_type=goss, importance_type=split, learning_rate=0.4,
min_split_gain=0.79, n_estimators=1000;, score=0.970 total time= 0.0s
[CV 2/5] END boosting_type=goss, importance_type=split, learning_rate=0.4,
min_split_gain=0.79, n_estimators=1000;, score=0.970 total time= 0.2s
[CV 3/5] END boosting_type=goss, importance_type=split, learning_rate=0.4,
min_split_gain=0.79, n_estimators=1000;, score=0.967 total time= 0.1s
[CV 4/5] END boosting_type=goss, importance_type=split, learning_rate=0.4,
min_split_gain=0.79, n_estimators=1000;, score=0.969 total time= 0.2s
[CV 5/5] END boosting_type=goss, importance_type=split, learning_rate=0.4,
min_split_gain=0.79, n_estimators=1000;, score=0.971 total time= 0.2s
[CV 1/5] END boosting_type=goss, importance_type=gain, learning_rate=0.2,
min_split_gain=0.79, n_estimators=500;, score=0.969 total time= 0.0s
[CV 2/5] END boosting_type=goss, importance_type=gain, learning_rate=0.2,
min_split_gain=0.79, n_estimators=500;, score=0.972 total time= 0.1s
[CV 3/5] END boosting_type=goss, importance_type=gain, learning_rate=0.2,
min_split_gain=0.79, n_estimators=500;, score=0.963 total time= 0.0s
[CV 4/5] END boosting_type=goss, importance_type=gain, learning_rate=0.2,
min_split_gain=0.79, n_estimators=500;, score=0.970 total time= 0.0s

```

```

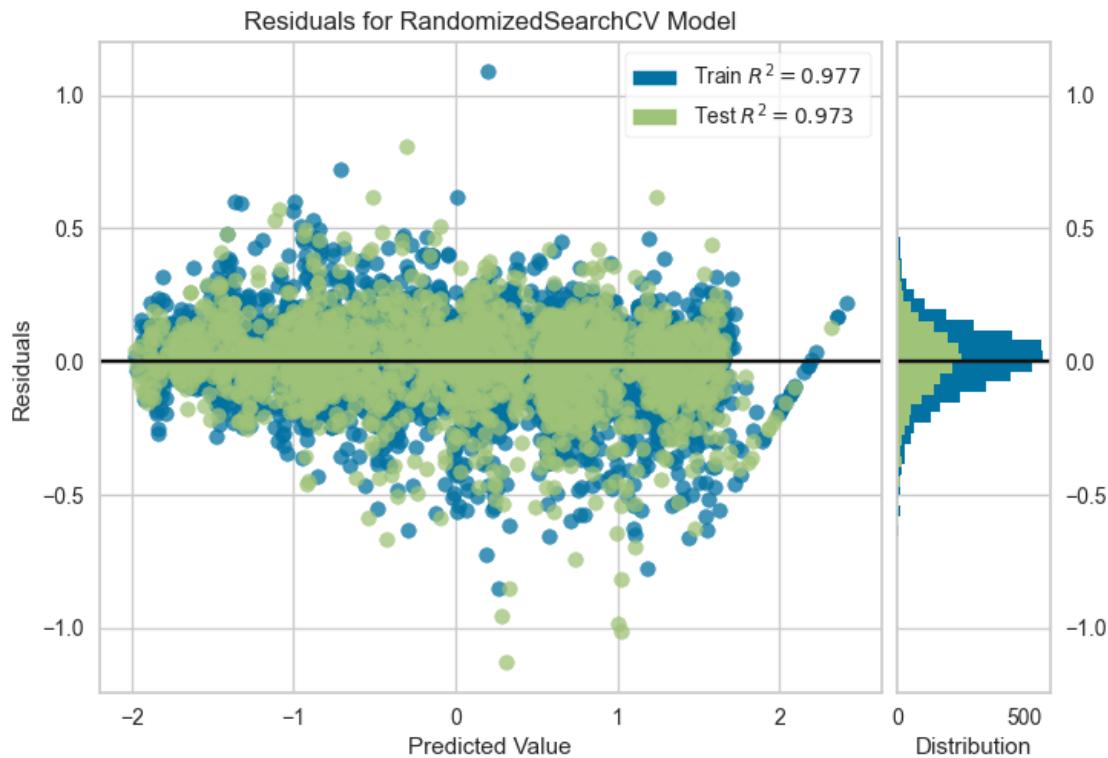
[CV 5/5] END boosting_type=goss, importance_type=gain, learning_rate=0.2,
min_split_gain=0.79, n_estimators=500;, score=0.972 total time= 0.0s
[CV 1/5] END boosting_type=goss, importance_type=split,
learning_rate=0.6000000000000001, min_split_gain=1, n_estimators=800;,
score=0.965 total time= 0.0s
[CV 2/5] END boosting_type=goss, importance_type=split,
learning_rate=0.6000000000000001, min_split_gain=1, n_estimators=800;,
score=0.965 total time= 0.0s
[CV 3/5] END boosting_type=goss, importance_type=split,
learning_rate=0.6000000000000001, min_split_gain=1, n_estimators=800;,
score=0.963 total time= 0.0s
[CV 4/5] END boosting_type=goss, importance_type=split,
learning_rate=0.6000000000000001, min_split_gain=1, n_estimators=800;,
score=0.964 total time= 0.0s
[CV 5/5] END boosting_type=goss, importance_type=split,
learning_rate=0.6000000000000001, min_split_gain=1, n_estimators=800;,
score=0.966 total time= 0.0s
[CV 1/5] END boosting_type=dart, importance_type=split, learning_rate=0.8,
min_split_gain=0.87, n_estimators=1000;, score=0.964 total time= 0.9s
[CV 2/5] END boosting_type=dart, importance_type=split, learning_rate=0.8,
min_split_gain=0.87, n_estimators=1000;, score=0.971 total time= 1.9s
[CV 3/5] END boosting_type=dart, importance_type=split, learning_rate=0.8,
min_split_gain=0.87, n_estimators=1000;, score=0.967 total time= 2.1s
[CV 4/5] END boosting_type=dart, importance_type=split, learning_rate=0.8,
min_split_gain=0.87, n_estimators=1000;, score=0.967 total time= 2.0s
[CV 5/5] END boosting_type=dart, importance_type=split, learning_rate=0.8,
min_split_gain=0.87, n_estimators=1000;, score=0.971 total time= 2.1s
[CV 1/5] END boosting_type=rf, importance_type=gain, learning_rate=0.8,
min_split_gain=0.79, n_estimators=800;, score=nan total time= 0.0s
[CV 2/5] END boosting_type=rf, importance_type=gain, learning_rate=0.8,
min_split_gain=0.79, n_estimators=800;, score=nan total time= 0.0s
[CV 3/5] END boosting_type=rf, importance_type=gain, learning_rate=0.8,
min_split_gain=0.79, n_estimators=800;, score=nan total time= 0.0s
[CV 4/5] END boosting_type=rf, importance_type=gain, learning_rate=0.8,
min_split_gain=0.79, n_estimators=800;, score=nan total time= 0.0s
[CV 5/5] END boosting_type=rf, importance_type=gain, learning_rate=0.8,
min_split_gain=0.79, n_estimators=800;, score=nan total time= 0.0s
[CV 1/5] END boosting_type=gbdt, importance_type=gain, learning_rate=0.4,
min_split_gain=0.87, n_estimators=800;, score=0.956 total time= 0.0s
[CV 2/5] END boosting_type=gbdt, importance_type=gain, learning_rate=0.4,
min_split_gain=0.87, n_estimators=800;, score=0.964 total time= 0.0s
[CV 3/5] END boosting_type=gbdt, importance_type=gain, learning_rate=0.4,
min_split_gain=0.87, n_estimators=800;, score=0.957 total time= 0.0s
[CV 4/5] END boosting_type=gbdt, importance_type=gain, learning_rate=0.4,
min_split_gain=0.87, n_estimators=800;, score=0.964 total time= 0.0s
[CV 5/5] END boosting_type=gbdt, importance_type=gain, learning_rate=0.4,
min_split_gain=0.87, n_estimators=800;, score=0.968 total time= 0.0s
[CV 1/5] END boosting_type=gbdt, importance_type=split, learning_rate=0.4,

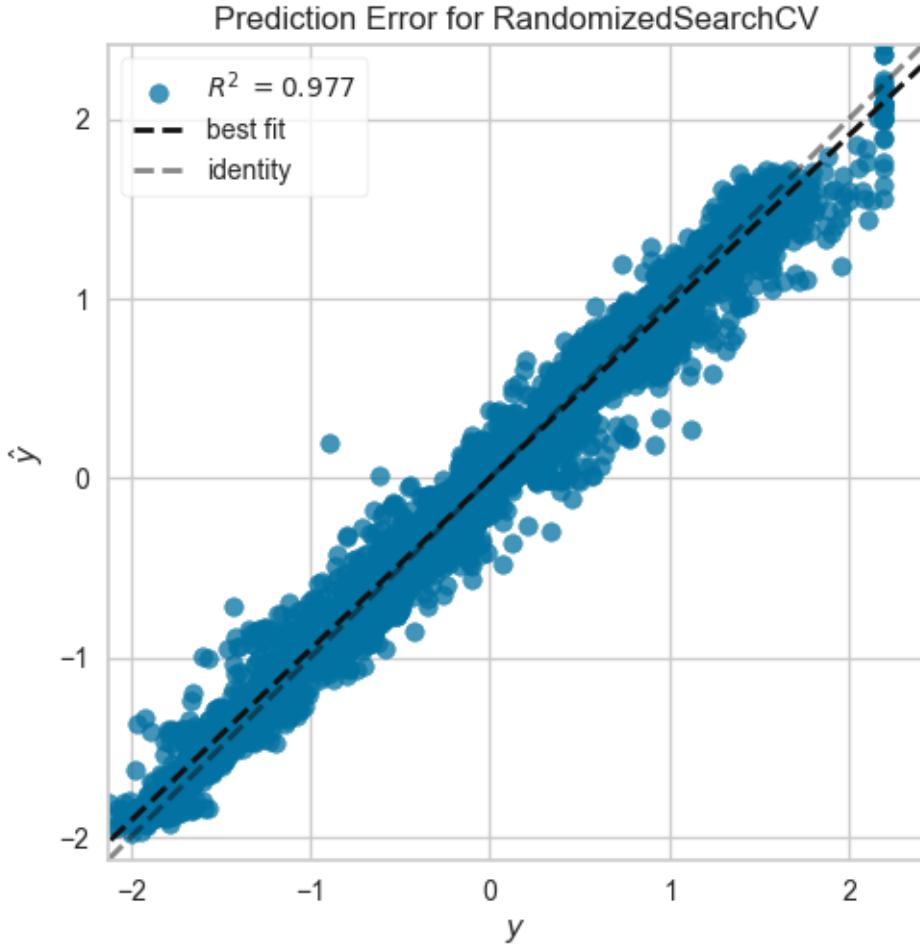
```

```

min_split_gain=0.87, n_estimators=100;, score=0.956 total time= 0.0s
[CV 2/5] END boosting_type=gbdt, importance_type=split, learning_rate=0.4,
min_split_gain=0.87, n_estimators=100;, score=0.964 total time= 0.0s
[CV 3/5] END boosting_type=gbdt, importance_type=split, learning_rate=0.4,
min_split_gain=0.87, n_estimators=100;, score=0.957 total time= 0.0s
[CV 4/5] END boosting_type=gbdt, importance_type=split, learning_rate=0.4,
min_split_gain=0.87, n_estimators=100;, score=0.964 total time= 0.0s
[CV 5/5] END boosting_type=gbdt, importance_type=split, learning_rate=0.4,
min_split_gain=0.87, n_estimators=100;, score=0.968 total time= 0.0s
Mean Absolute Percentage Error(MAPE): 0.8123791708530981
Root Mean Squared Error(RMSE): 0.1660202071635469
R2 Score: 0.9730744529902002

```





```
[108]: grid_lgbm.best_params_
```

```
[108]: {'n_estimators': 1000,
      'min_split_gain': 0.79,
      'learning_rate': 0.4,
      'importance_type': 'split',
      'boosting_type': 'goss'}
```

```
[109]: param_grid = {'n_estimators': [100,300,500,800,1000],
                  'criterion': ["squared_error", "absolute_error", "friedman_mse", "poisson"],
                  'max_features': ['auto','sqrt','log2'],
                  'bootstrap': [True,False],
                  'oob_score': [True,False],
                  'max_samples': [0.2,0.45,0.7,0.95]
                 }
```

```
grid_et = RandomizedSearchCV(ExtraTreesRegressor(), param_grid, verbose=2, cv=5)
train_and_evaluate_model(grid_et)
```

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

[CV] END bootstrap=True, criterion=squared\_error, max\_features=log2, max\_samples=0.2, n\_estimators=100, oob\_score=True; total time= 0.6s

[CV] END bootstrap=True, criterion=squared\_error, max\_features=log2, max\_samples=0.2, n\_estimators=100, oob\_score=True; total time= 0.6s

[CV] END bootstrap=True, criterion=squared\_error, max\_features=log2, max\_samples=0.2, n\_estimators=100, oob\_score=True; total time= 0.6s

[CV] END bootstrap=True, criterion=squared\_error, max\_features=log2, max\_samples=0.2, n\_estimators=100, oob\_score=True; total time= 0.2s

[CV] END bootstrap=True, criterion=squared\_error, max\_features=log2, max\_samples=0.2, n\_estimators=100, oob\_score=True; total time= 0.2s

[CV] END bootstrap=False, criterion=squared\_error, max\_features=auto, max\_samples=0.45, n\_estimators=1000, oob\_score=False; total time= 0.0s

[CV] END bootstrap=False, criterion=squared\_error, max\_features=auto, max\_samples=0.45, n\_estimators=1000, oob\_score=False; total time= 0.0s

[CV] END bootstrap=False, criterion=squared\_error, max\_features=auto, max\_samples=0.45, n\_estimators=1000, oob\_score=False; total time= 0.0s

[CV] END bootstrap=False, criterion=squared\_error, max\_features=auto, max\_samples=0.45, n\_estimators=1000, oob\_score=False; total time= 0.0s

[CV] END bootstrap=False, criterion=squared\_error, max\_features=auto, max\_samples=0.45, n\_estimators=1000, oob\_score=False; total time= 0.0s

[CV] END bootstrap=False, criterion=friedman\_mse, max\_features=log2, max\_samples=0.95, n\_estimators=1000, oob\_score=False; total time= 0.0s

[CV] END bootstrap=False, criterion=friedman\_mse, max\_features=log2, max\_samples=0.95, n\_estimators=1000, oob\_score=False; total time= 0.0s

[CV] END bootstrap=False, criterion=friedman\_mse, max\_features=log2, max\_samples=0.95, n\_estimators=1000, oob\_score=False; total time= 0.0s

[CV] END bootstrap=False, criterion=friedman\_mse, max\_features=log2, max\_samples=0.95, n\_estimators=1000, oob\_score=False; total time= 0.0s

[CV] END bootstrap=False, criterion=friedman\_mse, max\_features=log2, max\_samples=0.95, n\_estimators=1000, oob\_score=False; total time= 0.0s

[CV] END bootstrap=True, criterion=absolute\_error, max\_features=auto, max\_samples=0.2, n\_estimators=800, oob\_score=False; total time= 9.2s

[CV] END bootstrap=True, criterion=absolute\_error, max\_features=auto, max\_samples=0.2, n\_estimators=800, oob\_score=False; total time= 10.0s

[CV] END bootstrap=True, criterion=absolute\_error, max\_features=auto, max\_samples=0.2, n\_estimators=800, oob\_score=False; total time= 9.2s

[CV] END bootstrap=True, criterion=absolute\_error, max\_features=auto, max\_samples=0.2, n\_estimators=800, oob\_score=False; total time= 10.2s

[CV] END bootstrap=True, criterion=absolute\_error, max\_features=auto, max\_samples=0.2, n\_estimators=800, oob\_score=False; total time= 9.8s

[CV] END bootstrap=False, criterion=poisson, max\_features=auto, max\_samples=0.45, n\_estimators=800, oob\_score=True; total time= 0.0s

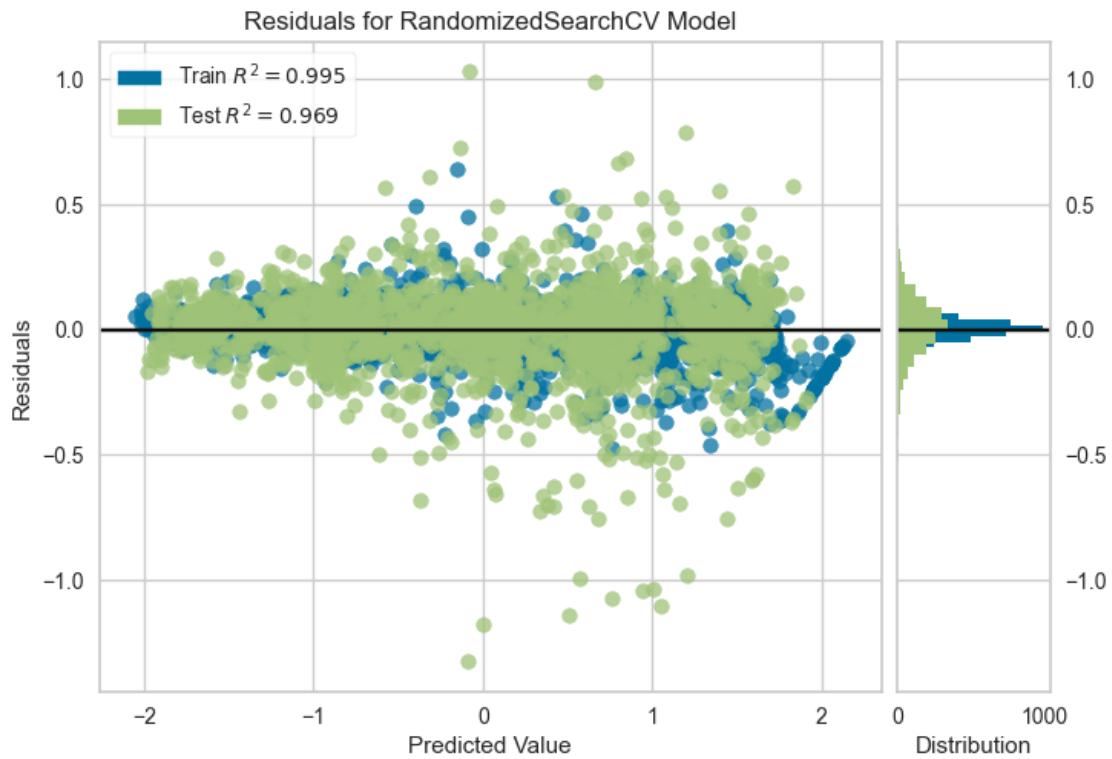
[CV] END bootstrap=False, criterion=poisson, max\_features=auto, max\_samples=0.45, n\_estimators=800, oob\_score=True; total time= 0.0s

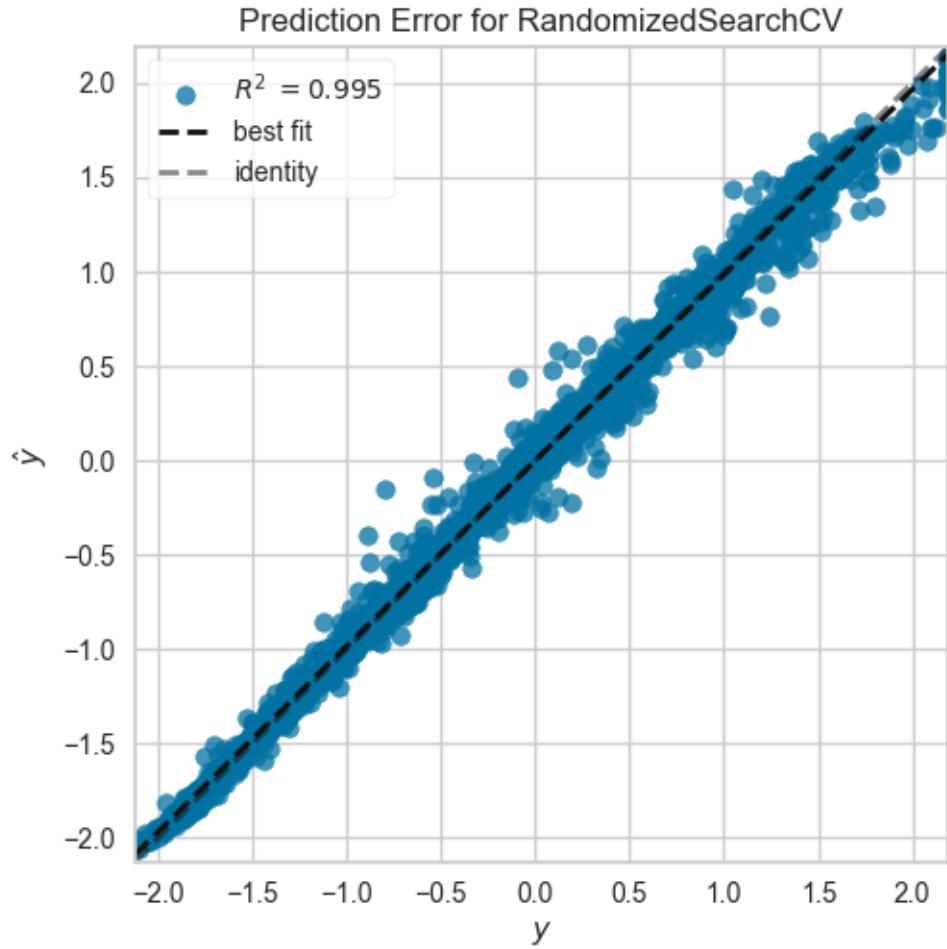


```

[CV] END bootstrap=True, criterion=poisson, max_features=log2, max_samples=0.95,
n_estimators=500, oob_score=False; total time= 0.0s
[CV] END bootstrap=True, criterion=poisson, max_features=log2, max_samples=0.95,
n_estimators=500, oob_score=False; total time= 0.0s
[CV] END bootstrap=True, criterion=poisson, max_features=log2, max_samples=0.95,
n_estimators=500, oob_score=False; total time= 0.0s
[CV] END bootstrap=True, criterion=poisson, max_features=log2, max_samples=0.95,
n_estimators=500, oob_score=False; total time= 0.0s
Mean Absolute Percentage Error(MAPE): 0.6428374445512449
Root Mean Squared Error(RMSE): 0.17914616108779607
R2 Score: 0.9686485490822321

```





```
[110]: grid_et.best_params_
```

```
[110]: {'oob_score': True,
      'n_estimators': 100,
      'max_samples': 0.95,
      'max_features': 'auto',
      'criterion': 'squared_error',
      'bootstrap': True}
```

```
[111]: param_grid = {'n_estimators': [100,400,700,900,1000],
                  'grow_policy': [0,1],
                  'learning_rate': [0.1,0.4,0.6,0.8,1],
                  'booster': ['gbtree','gblinear','dart'],
                  'sampling_method': ['uniform','gradient_based'],
                  'importance_type': [
                     'gain','weight','cover','total_gain','total_cover']
                 }
```

```

grid_xgb = RandomizedSearchCV(XGBRegressor(), param_grid, verbose=2, cv=5)
train_and_evaluate_model(grid_xgb)

```

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

[CV] END booster=gbtree, grow\_policy=0, importance\_type=gain, learning\_rate=1, n\_estimators=700, sampling\_method=uniform; total time= 0.0s

[CV] END booster=gbtree, grow\_policy=0, importance\_type=gain, learning\_rate=1, n\_estimators=700, sampling\_method=uniform; total time= 0.0s

[CV] END booster=gbtree, grow\_policy=0, importance\_type=gain, learning\_rate=1, n\_estimators=700, sampling\_method=uniform; total time= 0.0s

[CV] END booster=gbtree, grow\_policy=0, importance\_type=gain, learning\_rate=1, n\_estimators=700, sampling\_method=uniform; total time= 0.0s

[CV] END booster=gbtree, grow\_policy=0, importance\_type=gain, learning\_rate=1, n\_estimators=700, sampling\_method=uniform; total time= 0.0s

[CV] END booster=dart, grow\_policy=0, importance\_type=gain, learning\_rate=0.6, n\_estimators=700, sampling\_method=gradient\_based; total time= 0.0s

[CV] END booster=dart, grow\_policy=0, importance\_type=gain, learning\_rate=0.6, n\_estimators=700, sampling\_method=gradient\_based; total time= 0.0s

[CV] END booster=dart, grow\_policy=0, importance\_type=gain, learning\_rate=0.6, n\_estimators=700, sampling\_method=gradient\_based; total time= 0.0s

[CV] END booster=dart, grow\_policy=0, importance\_type=gain, learning\_rate=0.6, n\_estimators=700, sampling\_method=gradient\_based; total time= 0.0s

[CV] END booster=dart, grow\_policy=0, importance\_type=gain, learning\_rate=0.6, n\_estimators=700, sampling\_method=gradient\_based; total time= 0.0s

[CV] END booster=dart, grow\_policy=0, importance\_type=gain, learning\_rate=0.6, n\_estimators=700, sampling\_method=gradient\_based; total time= 0.0s

[CV] END booster=gbtree, grow\_policy=1, importance\_type=gain, learning\_rate=0.6, n\_estimators=1000, sampling\_method=gradient\_based; total time= 0.0s

[CV] END booster=gbtree, grow\_policy=1, importance\_type=gain, learning\_rate=0.6, n\_estimators=1000, sampling\_method=gradient\_based; total time= 0.0s

[CV] END booster=gbtree, grow\_policy=1, importance\_type=gain, learning\_rate=0.6, n\_estimators=1000, sampling\_method=gradient\_based; total time= 0.0s

[CV] END booster=gbtree, grow\_policy=1, importance\_type=gain, learning\_rate=0.6, n\_estimators=1000, sampling\_method=gradient\_based; total time= 0.0s

[CV] END booster=gbtree, grow\_policy=1, importance\_type=gain, learning\_rate=0.6, n\_estimators=1000, sampling\_method=gradient\_based; total time= 0.0s

[CV] END booster=gbtree, grow\_policy=1, importance\_type=gain, learning\_rate=0.6, n\_estimators=1000, sampling\_method=gradient\_based; total time= 0.0s

[CV] END booster=gbtree, grow\_policy=1, importance\_type=gain, learning\_rate=0.6, n\_estimators=1000, sampling\_method=gradient\_based; total time= 0.0s

[CV] END booster=gbtree, grow\_policy=0, importance\_type=weight, learning\_rate=1, n\_estimators=100, sampling\_method=uniform; total time= 0.0s

[CV] END booster=gbtree, grow\_policy=0, importance\_type=weight, learning\_rate=1, n\_estimators=100, sampling\_method=uniform; total time= 0.0s

[CV] END booster=gbtree, grow\_policy=0, importance\_type=weight, learning\_rate=1, n\_estimators=100, sampling\_method=uniform; total time= 0.0s

[CV] END booster=gbtree, grow\_policy=0, importance\_type=weight, learning\_rate=1, n\_estimators=100, sampling\_method=uniform; total time= 0.0s

[CV] END booster=dart, grow\_policy=1, importance\_type=total\_cover, learning\_rate=0.1, n\_estimators=700, sampling\_method=uniform; total time= 0.0s

[CV] END booster=dart, grow\_policy=1, importance\_type=total\_cover,

```

learning_rate=0.1, n_estimators=700, sampling_method=uniform; total time= 0.0s
[CV] END booster=dart, grow_policy=1, importance_type=total_cover,
learning_rate=0.1, n_estimators=700, sampling_method=uniform; total time= 0.0s
[CV] END booster=dart, grow_policy=1, importance_type=total_cover,
learning_rate=0.1, n_estimators=700, sampling_method=uniform; total time= 0.0s
[CV] END booster=dart, grow_policy=1, importance_type=total_cover,
learning_rate=0.1, n_estimators=700, sampling_method=uniform; total time= 0.0s
[CV] END booster=dart, grow_policy=1, importance_type=total_gain,
learning_rate=1, n_estimators=900, sampling_method=uniform; total time= 0.0s
[CV] END booster=dart, grow_policy=1, importance_type=total_gain,
learning_rate=1, n_estimators=900, sampling_method=uniform; total time= 0.0s
[CV] END booster=dart, grow_policy=1, importance_type=total_gain,
learning_rate=1, n_estimators=900, sampling_method=uniform; total time= 0.0s
[CV] END booster=dart, grow_policy=1, importance_type=total_gain,
learning_rate=1, n_estimators=900, sampling_method=uniform; total time= 0.0s
[CV] END booster=dart, grow_policy=1, importance_type=total_gain,
learning_rate=1, n_estimators=900, sampling_method=uniform; total time= 0.0s
[CV] END booster=dart, grow_policy=1, importance_type=total_gain,
learning_rate=1, n_estimators=900, sampling_method=uniform; total time= 0.0s
[CV] END booster=dart, grow_policy=1, importance_type=gain, learning_rate=0.6,
n_estimators=1000, sampling_method=gradient_based; total time= 0.0s
[CV] END booster=dart, grow_policy=1, importance_type=gain, learning_rate=0.6,
n_estimators=1000, sampling_method=gradient_based; total time= 0.0s
[CV] END booster=dart, grow_policy=1, importance_type=gain, learning_rate=0.6,
n_estimators=1000, sampling_method=gradient_based; total time= 0.0s
[CV] END booster=dart, grow_policy=1, importance_type=gain, learning_rate=0.6,
n_estimators=1000, sampling_method=gradient_based; total time= 0.0s
[CV] END booster=dart, grow_policy=1, importance_type=gain, learning_rate=0.6,
n_estimators=1000, sampling_method=gradient_based; total time= 0.0s
[01:02:53] WARNING: C:\Users\dev-admin\croot2\xgboost-
split_1675461376218\work\src\learner.cc:767:
Parameters: { "grow_policy", "sampling_method" } are not used.

[CV] END booster=gblinear, grow_policy=1, importance_type=gain,
learning_rate=0.8, n_estimators=100, sampling_method=uniform; total time= 0.0s
[01:02:53] WARNING: C:\Users\dev-admin\croot2\xgboost-
split_1675461376218\work\src\learner.cc:767:
Parameters: { "grow_policy", "sampling_method" } are not used.

[CV] END booster=gblinear, grow_policy=1, importance_type=gain,
learning_rate=0.8, n_estimators=100, sampling_method=uniform; total time= 0.0s
[01:02:53] WARNING: C:\Users\dev-admin\croot2\xgboost-
split_1675461376218\work\src\learner.cc:767:
Parameters: { "grow_policy", "sampling_method" } are not used.

[CV] END booster=gblinear, grow_policy=1, importance_type=gain,
learning_rate=0.8, n_estimators=100, sampling_method=uniform; total time= 0.0s
[01:02:53] WARNING: C:\Users\dev-admin\croot2\xgboost-
split_1675461376218\work\src\learner.cc:767:
Parameters: { "grow_policy", "sampling_method" } are not used.

```

```
[CV] END booster=gblinear, grow_policy=1, importance_type=gain,
learning_rate=0.8, n_estimators=100, sampling_method=uniform; total time= 0.0s
[01:02:53] WARNING: C:\Users\dev-admin\croot2\xgboost-
split_1675461376218\work\src\learner.cc:767:
Parameters: { "grow_policy", "sampling_method" } are not used.

[CV] END booster=gblinear, grow_policy=1, importance_type=gain,
learning_rate=0.8, n_estimators=100, sampling_method=uniform; total time= 0.0s
[01:02:53] WARNING: C:\Users\dev-admin\croot2\xgboost-
split_1675461376218\work\src\learner.cc:767:
Parameters: { "grow_policy", "sampling_method" } are not used.

[CV] END booster=gblinear, grow_policy=0, importance_type=cover,
learning_rate=0.4, n_estimators=1000, sampling_method=uniform; total time=
0.1s
[01:02:54] WARNING: C:\Users\dev-admin\croot2\xgboost-
split_1675461376218\work\src\learner.cc:767:
Parameters: { "grow_policy", "sampling_method" } are not used.

[CV] END booster=gblinear, grow_policy=0, importance_type=cover,
learning_rate=0.4, n_estimators=1000, sampling_method=uniform; total time=
0.1s
[01:02:54] WARNING: C:\Users\dev-admin\croot2\xgboost-
split_1675461376218\work\src\learner.cc:767:
Parameters: { "grow_policy", "sampling_method" } are not used.

[CV] END booster=gblinear, grow_policy=0, importance_type=cover,
learning_rate=0.4, n_estimators=1000, sampling_method=uniform; total time=
0.1s
[01:02:54] WARNING: C:\Users\dev-admin\croot2\xgboost-
split_1675461376218\work\src\learner.cc:767:
Parameters: { "grow_policy", "sampling_method" } are not used.

[CV] END booster=gblinear, grow_policy=0, importance_type=cover,
learning_rate=0.4, n_estimators=1000, sampling_method=uniform; total time=
0.1s
[01:02:54] WARNING: C:\Users\dev-admin\croot2\xgboost-
split_1675461376218\work\src\learner.cc:767:
Parameters: { "grow_policy", "sampling_method" } are not used.

[CV] END booster=gbtree, grow_policy=1, importance_type=weight,
learning_rate=0.6, n_estimators=400, sampling_method=gradient_based; total time=
0.0s
[CV] END booster=gbtree, grow_policy=1, importance_type=weight,
```

```

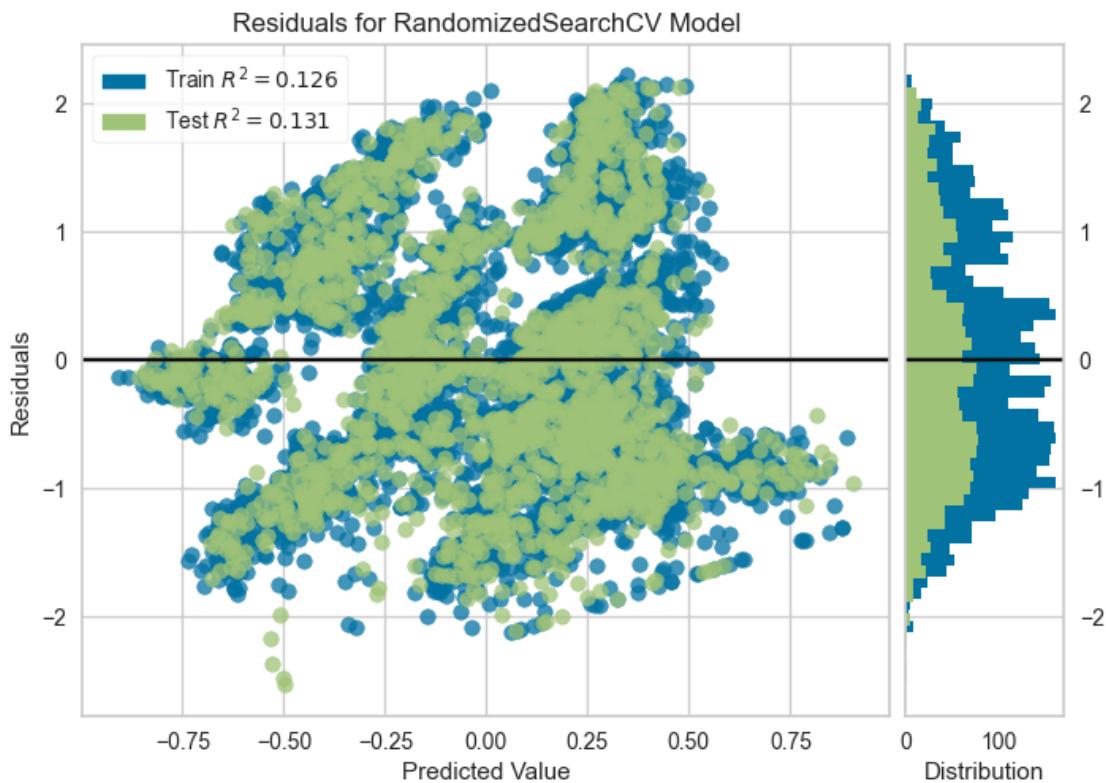
learning_rate=0.6, n_estimators=400, sampling_method=gradient_based; total time=
0.0s
[CV] END booster=gbtree, grow_policy=1, importance_type=weight,
learning_rate=0.6, n_estimators=400, sampling_method=gradient_based; total time=
0.0s
[CV] END booster=gbtree, grow_policy=1, importance_type=weight,
learning_rate=0.6, n_estimators=400, sampling_method=gradient_based; total time=
0.0s
[CV] END booster=gbtree, grow_policy=1, importance_type=weight,
learning_rate=0.6, n_estimators=400, sampling_method=gradient_based; total time=
0.0s
[01:02:54] WARNING: C:\Users\dev-admin\croot2\xgboost-
split_1675461376218\work\src\learner.cc:767:
Parameters: { "grow_policy", "sampling_method" } are not used.

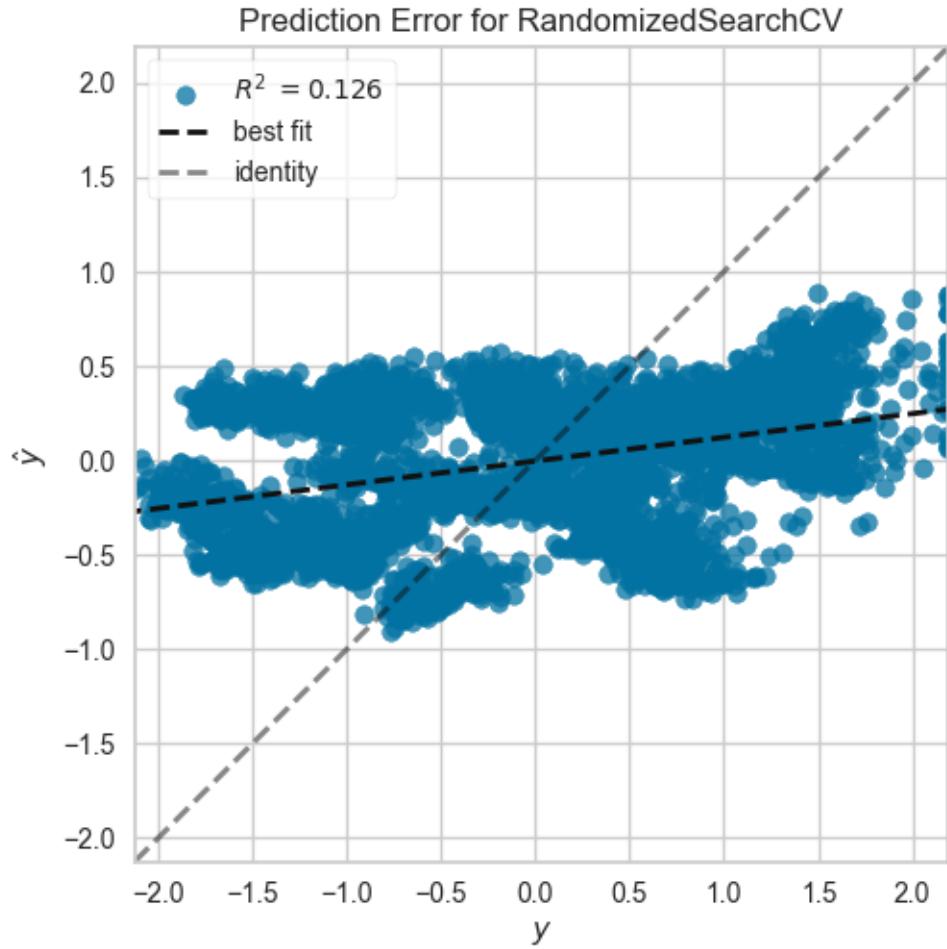
```

Mean Absolute Percentage Error(MAPE): 1.8642096254763194

Root Mean Squared Error(RMSE): 0.9429977568683355

R2 Score: 0.1313117414852658





```
[112]: grid_xgb.best_params_
```

```
[112]: {'sampling_method': 'uniform',
      'n_estimators': 100,
      'learning_rate': 0.8,
      'importance_type': 'gain',
      'grow_policy': 1,
      'booster': 'gblinear'}
```

```
[113]: param_grid = {'loss': [
    'squared_error', 'huber', 'epsilon_insensitive', 'squared_epsilon_insensitive'],
    'penalty': ['l2', 'l1', 'elasticnet'],
    'l1_ratio': [0.15, 0.45, 0.68, 0.81, 0.97],
    'alpha': [0.0001, 0.001, 0.01, 0.1, 1],
    'shuffle': [True, False],
    'learning_rate': ['adaptive', 'constant', 'optimal', 'invscaling'],
    'epsilon': np.linspace(0.001, 100, 10),
```

```

    'average': [True, False]
}

grid_sgd = RandomizedSearchCV(SGDRegressor(), param_grid, verbose=2, cv=5)
train_and_evaluate_model(grid_sgd)

Fitting 5 folds for each of 10 candidates, totalling 50 fits
[CV] END alpha=1, average=False, epsilon=11.11999999999998, l1_ratio=0.15,
learning_rate=adaptive, loss=squared_error, penalty=l2, shuffle=True; total
time= 0.0s
[CV] END alpha=1, average=False, epsilon=11.11999999999998, l1_ratio=0.15,
learning_rate=adaptive, loss=squared_error, penalty=l2, shuffle=True; total
time= 0.0s
[CV] END alpha=1, average=False, epsilon=11.11999999999998, l1_ratio=0.15,
learning_rate=adaptive, loss=squared_error, penalty=l2, shuffle=True; total
time= 0.0s
[CV] END alpha=1, average=False, epsilon=11.11999999999998, l1_ratio=0.15,
learning_rate=adaptive, loss=squared_error, penalty=l2, shuffle=True; total
time= 0.0s
[CV] END alpha=1, average=False, epsilon=11.11999999999998, l1_ratio=0.15,
learning_rate=adaptive, loss=squared_error, penalty=l2, shuffle=True; total
time= 0.0s
[CV] END alpha=1, average=False, epsilon=11.11999999999998, l1_ratio=0.15,
learning_rate=adaptive, loss=squared_error, penalty=l2, shuffle=True; total
time= 0.0s
[CV] END alpha=1, average=False, epsilon=11.11999999999998, l1_ratio=0.97,
learning_rate=optimal, loss=huber, penalty=l2, shuffle=False; total time= 0.0s
[CV] END alpha=1, average=False, epsilon=11.11999999999998, l1_ratio=0.97,
learning_rate=optimal, loss=huber, penalty=l2, shuffle=False; total time= 0.0s
[CV] END alpha=1, average=False, epsilon=11.11999999999998, l1_ratio=0.97,
learning_rate=optimal, loss=huber, penalty=l2, shuffle=False; total time= 0.0s
[CV] END alpha=1, average=False, epsilon=11.11999999999998, l1_ratio=0.97,
learning_rate=optimal, loss=huber, penalty=l2, shuffle=False; total time= 0.0s
[CV] END alpha=1, average=False, epsilon=11.11999999999998, l1_ratio=0.97,
learning_rate=optimal, loss=huber, penalty=l2, shuffle=False; total time= 0.0s
[CV] END alpha=1, average=False, epsilon=100.0, l1_ratio=0.97,
learning_rate=constant, loss=squared_epsilon_insensitive, penalty=l1,
shuffle=False; total time= 0.0s
[CV] END alpha=1, average=False, epsilon=100.0, l1_ratio=0.97,
learning_rate=constant, loss=squared_epsilon_insensitive, penalty=l1,
shuffle=False; total time= 0.0s
[CV] END alpha=1, average=False, epsilon=100.0, l1_ratio=0.97,
learning_rate=constant, loss=squared_epsilon_insensitive, penalty=l1,
shuffle=False; total time= 0.0s
[CV] END alpha=1, average=False, epsilon=100.0, l1_ratio=0.97,
learning_rate=constant, loss=squared_epsilon_insensitive, penalty=l1,
shuffle=False; total time= 0.0s
[CV] END alpha=0.01, average=False, epsilon=22.223, l1_ratio=0.15,

```

```

learning_rate=optimal, loss=squared_error, penalty=elasticnet, shuffle=False;
total time= 0.4s
[CV] END alpha=0.01, average=False, epsilon=22.223, l1_ratio=0.15,
learning_rate=optimal, loss=squared_error, penalty=elasticnet, shuffle=False;
total time= 0.2s
[CV] END alpha=0.01, average=False, epsilon=22.223, l1_ratio=0.15,
learning_rate=optimal, loss=squared_error, penalty=elasticnet, shuffle=False;
total time= 0.1s
[CV] END alpha=0.01, average=False, epsilon=22.223, l1_ratio=0.15,
learning_rate=optimal, loss=squared_error, penalty=elasticnet, shuffle=False;
total time= 0.1s
[CV] END alpha=0.01, average=False, epsilon=22.223, l1_ratio=0.15,
learning_rate=optimal, loss=squared_error, penalty=elasticnet, shuffle=False;
total time= 0.1s
[CV] END alpha=0.0001, average=False, epsilon=0.001, l1_ratio=0.81,
learning_rate=constant, loss=squared_epsilon_insensitive, penalty=elasticnet,
shuffle=True; total time= 0.0s
[CV] END alpha=0.0001, average=False, epsilon=0.001, l1_ratio=0.81,
learning_rate=constant, loss=squared_epsilon_insensitive, penalty=elasticnet,
shuffle=True; total time= 0.0s
[CV] END alpha=0.0001, average=False, epsilon=0.001, l1_ratio=0.81,
learning_rate=constant, loss=squared_epsilon_insensitive, penalty=elasticnet,
shuffle=True; total time= 0.0s
[CV] END alpha=0.0001, average=False, epsilon=0.001, l1_ratio=0.81,
learning_rate=constant, loss=squared_epsilon_insensitive, penalty=elasticnet,
shuffle=True; total time= 0.0s
[CV] END alpha=0.0001, average=False, epsilon=0.001, l1_ratio=0.81,
learning_rate=constant, loss=squared_epsilon_insensitive, penalty=elasticnet,
shuffle=True; total time= 0.0s
[CV] END alpha=0.0001, average=True, epsilon=100.0, l1_ratio=0.45,
learning_rate=invscaling, loss=squared_epsilon_insensitive, penalty=12,
shuffle=False; total time= 0.0s
[CV] END alpha=0.0001, average=True, epsilon=100.0, l1_ratio=0.45,
learning_rate=invscaling, loss=squared_epsilon_insensitive, penalty=12,
shuffle=False; total time= 0.0s
[CV] END alpha=0.0001, average=True, epsilon=100.0, l1_ratio=0.45,
learning_rate=invscaling, loss=squared_epsilon_insensitive, penalty=12,
shuffle=False; total time= 0.0s
[CV] END alpha=0.0001, average=True, epsilon=100.0, l1_ratio=0.45,
learning_rate=invscaling, loss=squared_epsilon_insensitive, penalty=12,
shuffle=False; total time= 0.0s
[CV] END alpha=0.0001, average=True, epsilon=100.0, l1_ratio=0.45,
learning_rate=invscaling, loss=squared_epsilon_insensitive, penalty=12,
shuffle=False; total time= 0.0s
[CV] END alpha=1, average=True, epsilon=88.889, l1_ratio=0.81,
learning_rate=invscaling, loss=epsilon_insensitive, penalty=11, shuffle=True;
total time= 0.0s
[CV] END alpha=1, average=True, epsilon=88.889, l1_ratio=0.81,

```

```

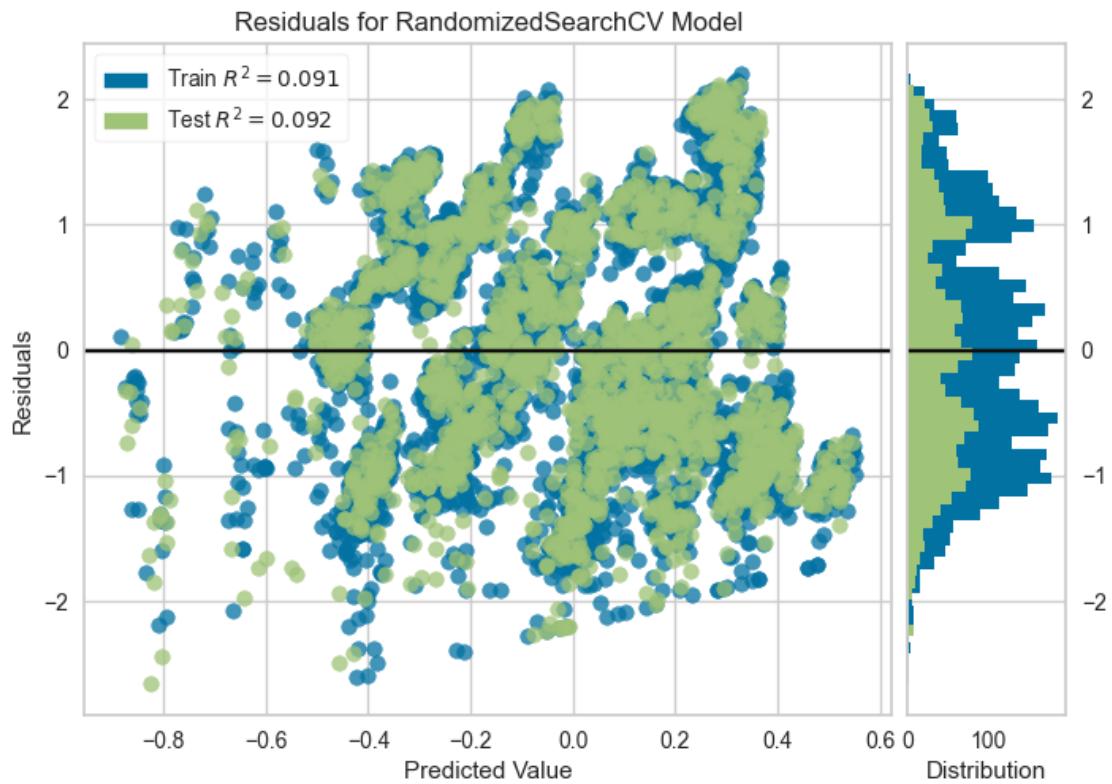
learning_rate=invscaling, loss=epsilon_insensitive, penalty=l1, shuffle=True;
total time= 0.0s
[CV] END alpha=1, average=True, epsilon=88.889, l1_ratio=0.81,
learning_rate=invscaling, loss=epsilon_insensitive, penalty=l1, shuffle=True;
total time= 0.0s
[CV] END alpha=1, average=True, epsilon=88.889, l1_ratio=0.81,
learning_rate=invscaling, loss=epsilon_insensitive, penalty=l1, shuffle=True;
total time= 0.0s
[CV] END alpha=1, average=True, epsilon=88.889, l1_ratio=0.81,
learning_rate=invscaling, loss=epsilon_insensitive, penalty=l1, shuffle=True;
total time= 0.0s
[CV] END alpha=0.1, average=True, epsilon=22.223, l1_ratio=0.68,
learning_rate=adaptive, loss=huber, penalty=l1, shuffle=False; total time=
0.0s
[CV] END alpha=0.1, average=True, epsilon=22.223, l1_ratio=0.68,
learning_rate=adaptive, loss=huber, penalty=l1, shuffle=False; total time=
0.0s
[CV] END alpha=0.1, average=True, epsilon=22.223, l1_ratio=0.68,
learning_rate=adaptive, loss=huber, penalty=l1, shuffle=False; total time=
0.0s
[CV] END alpha=0.1, average=True, epsilon=22.223, l1_ratio=0.68,
learning_rate=adaptive, loss=huber, penalty=l1, shuffle=False; total time=
0.0s
[CV] END alpha=0.1, average=True, epsilon=22.223, l1_ratio=0.68,
learning_rate=adaptive, loss=huber, penalty=l1, shuffle=False; total time=
0.0s
[CV] END alpha=0.001, average=True, epsilon=66.667, l1_ratio=0.15,
learning_rate=adaptive, loss=squared_error, penalty=l1, shuffle=False; total
time= 0.0s
[CV] END alpha=0.001, average=True, epsilon=66.667, l1_ratio=0.15,
learning_rate=adaptive, loss=squared_error, penalty=l1, shuffle=False; total
time= 0.0s
[CV] END alpha=0.001, average=True, epsilon=66.667, l1_ratio=0.15,
learning_rate=adaptive, loss=squared_error, penalty=l1, shuffle=False; total
time= 0.0s
[CV] END alpha=0.001, average=True, epsilon=66.667, l1_ratio=0.15,
learning_rate=adaptive, loss=squared_error, penalty=l1, shuffle=False; total
time= 0.0s
[CV] END alpha=0.1, average=True, epsilon=88.889, l1_ratio=0.97,
learning_rate=optimal, loss=squared_epsilon_insensitive, penalty=elasticnet,
shuffle=False; total time= 0.0s
[CV] END alpha=0.1, average=True, epsilon=88.889, l1_ratio=0.97,
learning_rate=optimal, loss=squared_epsilon_insensitive, penalty=elasticnet,
shuffle=False; total time= 0.0s
[CV] END alpha=0.1, average=True, epsilon=88.889, l1_ratio=0.97,

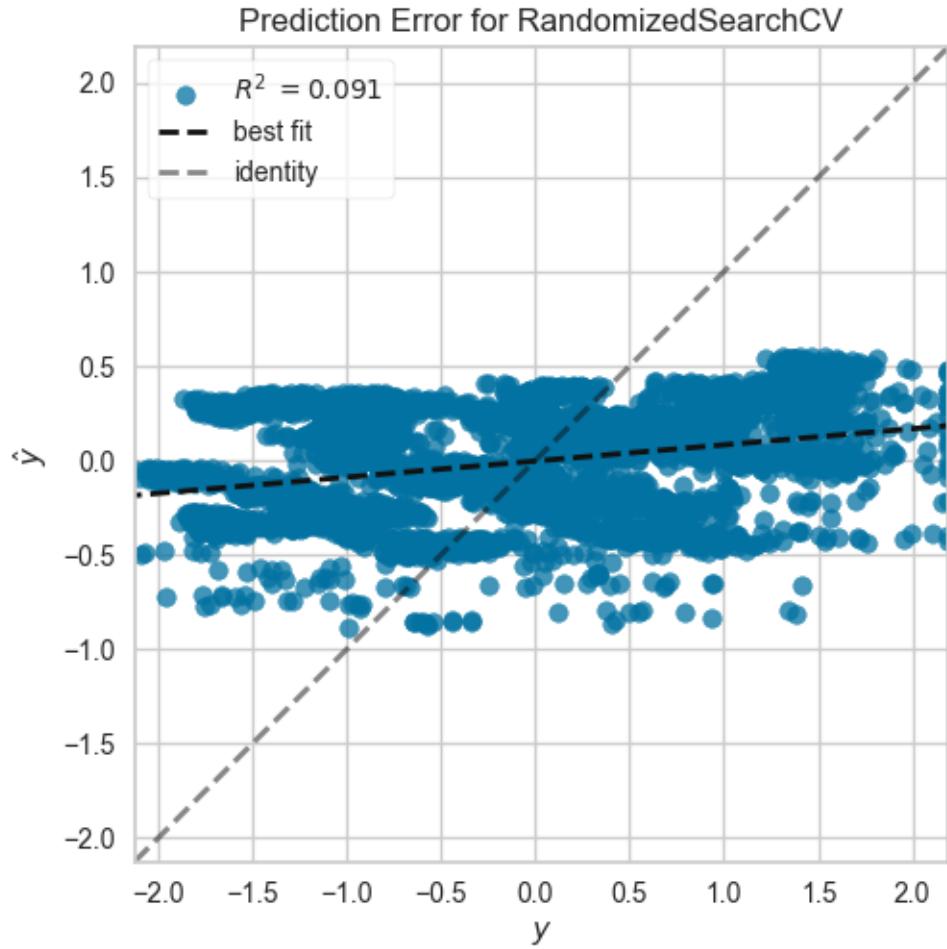
```

```

learning_rate=optimal, loss=squared_epsilon_insensitive, penalty=elasticnet,
shuffle=False; total time= 0.0s
[CV] END alpha=0.1, average=True, epsilon=88.889, l1_ratio=0.97,
learning_rate=optimal, loss=squared_epsilon_insensitive, penalty=elasticnet,
shuffle=False; total time= 0.0s
[CV] END alpha=0.1, average=True, epsilon=88.889, l1_ratio=0.97,
learning_rate=optimal, loss=squared_epsilon_insensitive, penalty=elasticnet,
shuffle=False; total time= 0.0s
Mean Absolute Percentage Error(MAPE): 2.2562796873375857
Root Mean Squared Error(RMSE): 0.964216413840828
R2 Score: 0.09177872177690394

```





```
[114]: grid_sgd.best_params_
```

```
[114]: {'shuffle': False,
      'penalty': 'l1',
      'loss': 'squared_error',
      'learning_rate': 'adaptive',
      'l1_ratio': 0.15,
      'epsilon': 66.667,
      'average': True,
      'alpha': 0.001}
```

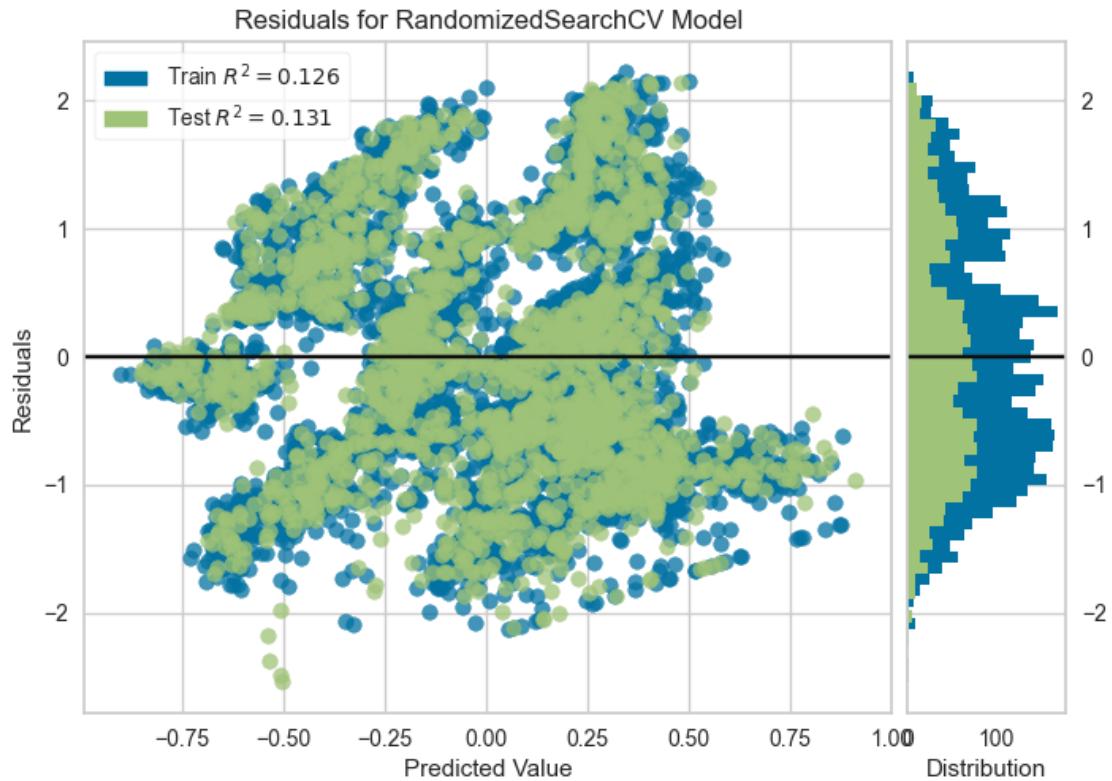
```
[115]: param_grid = {'epsilon': np.linspace(1,10,10),
                  'alpha': np.linspace(0.0001,10,10)}

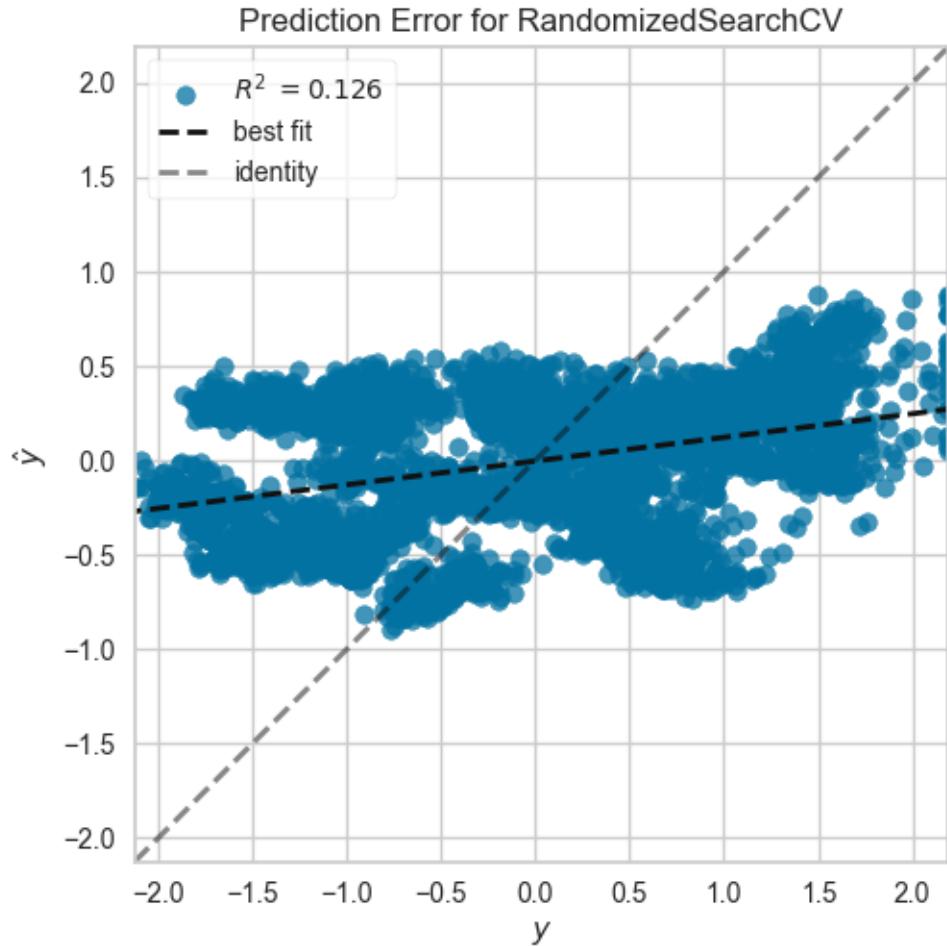
grid_huber = RandomizedSearchCV(HuberRegressor(), param_grid, verbose=2, cv=5)
train_and_evaluate_model(grid_huber)
```

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



```
[CV] END ...alpha=3.3334, epsilon=10.0; total time= 0.0s
[CV] END ...alpha=3.3334, epsilon=10.0; total time= 0.0s
Mean Absolute Percentage Error(MAPE): 1.8583990965205004
Root Mean Squared Error(RMSE): 0.9429069446854338
R2 Score: 0.1314790455492416
```





```
[116]: grid_hubер.best_params_
```

```
[116]: {'epsilon': 5.0, 'alpha': 10.0}
```

```
[117]: param_grid = {'link': ['auto', 'identity', 'log'],
                  'solver': ['lbfgs', 'newton-cholesky'],
                  'alpha': np.linspace(0.0001, 10, 10)}
```

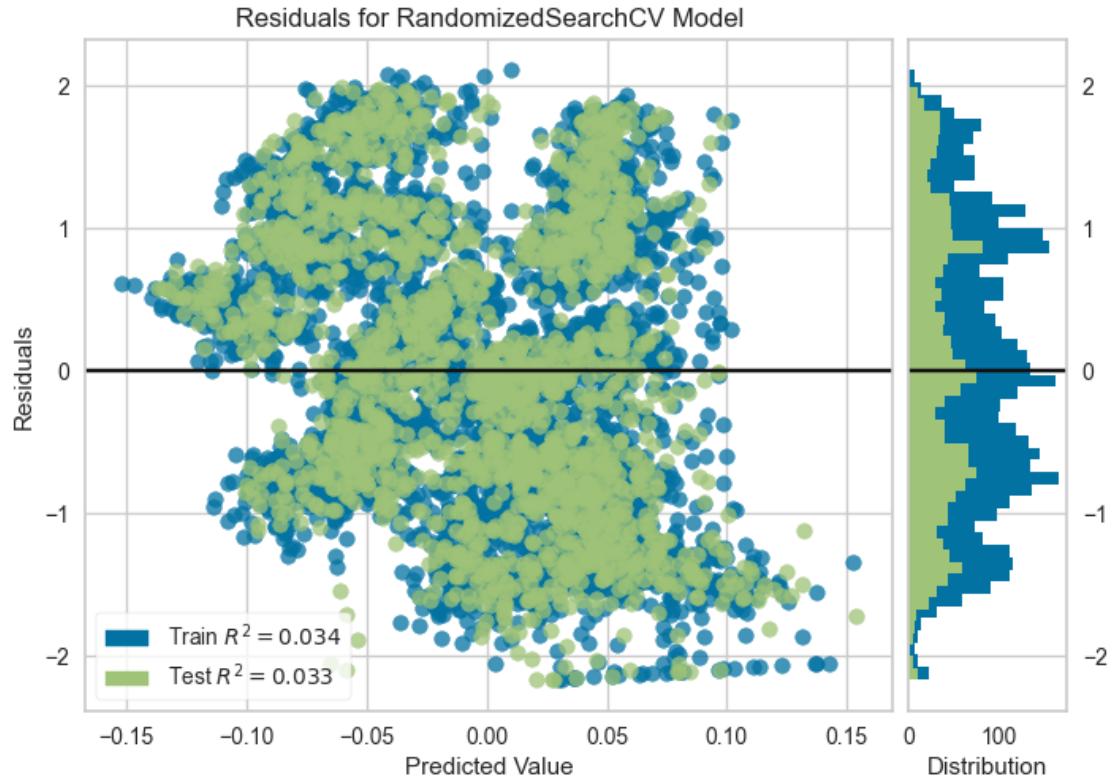
```
grid_tweedie = RandomizedSearchCV(TweedieRegressor(), param_grid, verbose=2, cv=5)
train_and_evaluate_model(grid_tweedie)
```

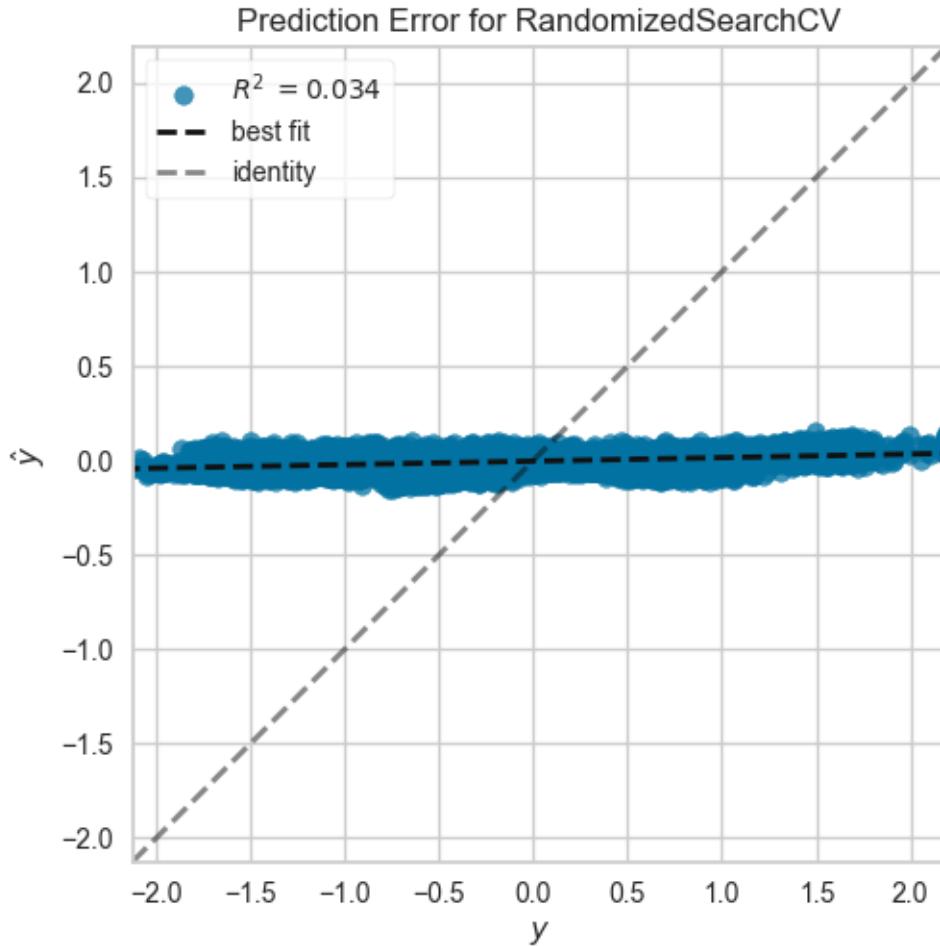
Fitting 5 folds for each of 10 candidates, totalling 50 fits

```
[CV] END ...alpha=10.0, link=identity, solver=lbfgs; total time= 0.0s
[CV] END ...alpha=1.1112, link=log, solver=lbfgs; total time= 0.0s
```



```
[CV] END ...alpha=8.8889, link=auto, solver=lbfgs; total time= 0.0s  
Mean Absolute Percentage Error(MAPE): 1.0537279069678445  
Root Mean Squared Error(RMSE): 0.9946842583367421  
R2 Score: 0.03347493298077597
```





```
[118]: grid_tweedie.best_params_
```

```
[118]: {'solver': 'newton-cholesky', 'link': 'identity', 'alpha': 5.5556}
```

```
[119]: param_grid = {'C': [0.0001, 0.001, 0.01, 0.1, 1, 10],
                  'loss': ['epsilon_insensitive', 'squared_epsilon_insensitive'],
                  'epsilon': np.linspace(0.001, 1, 5),
                  'shuffle': [True, False],
                  'average': [True, False]}

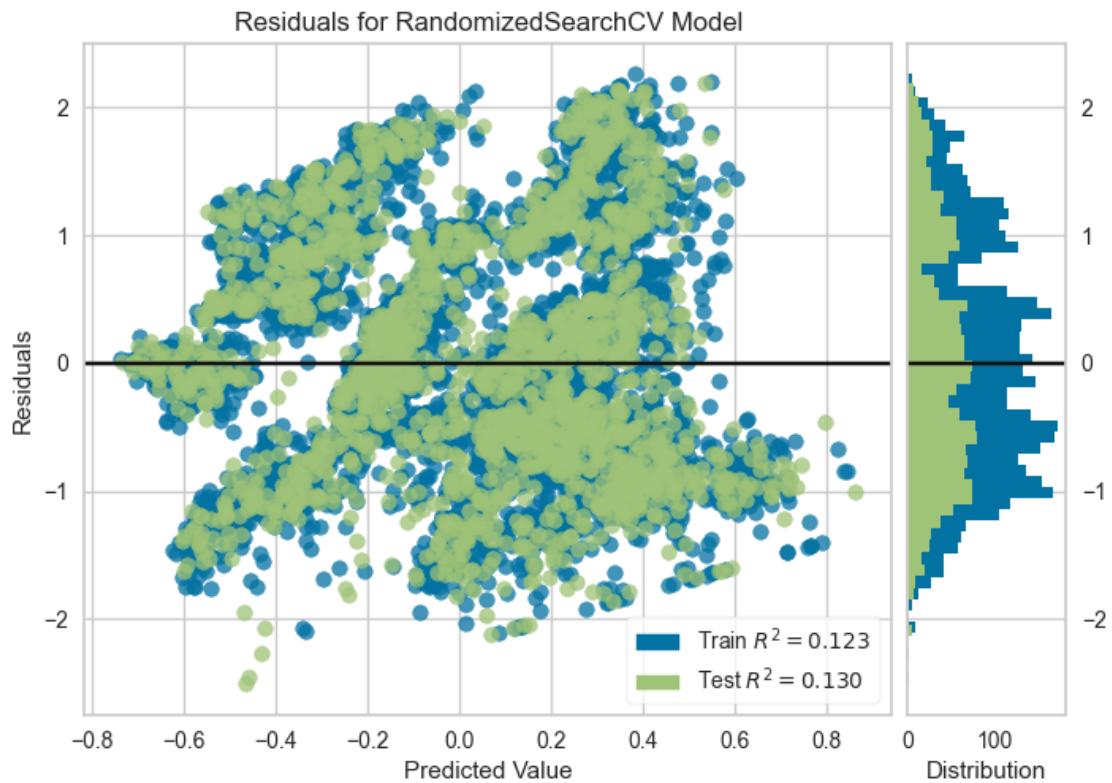
grid_pa = RandomizedSearchCV(PassiveAggressiveRegressor(), param_grid, verbose=2, cv=5)
train_and_evaluate_model(grid_pa)
```

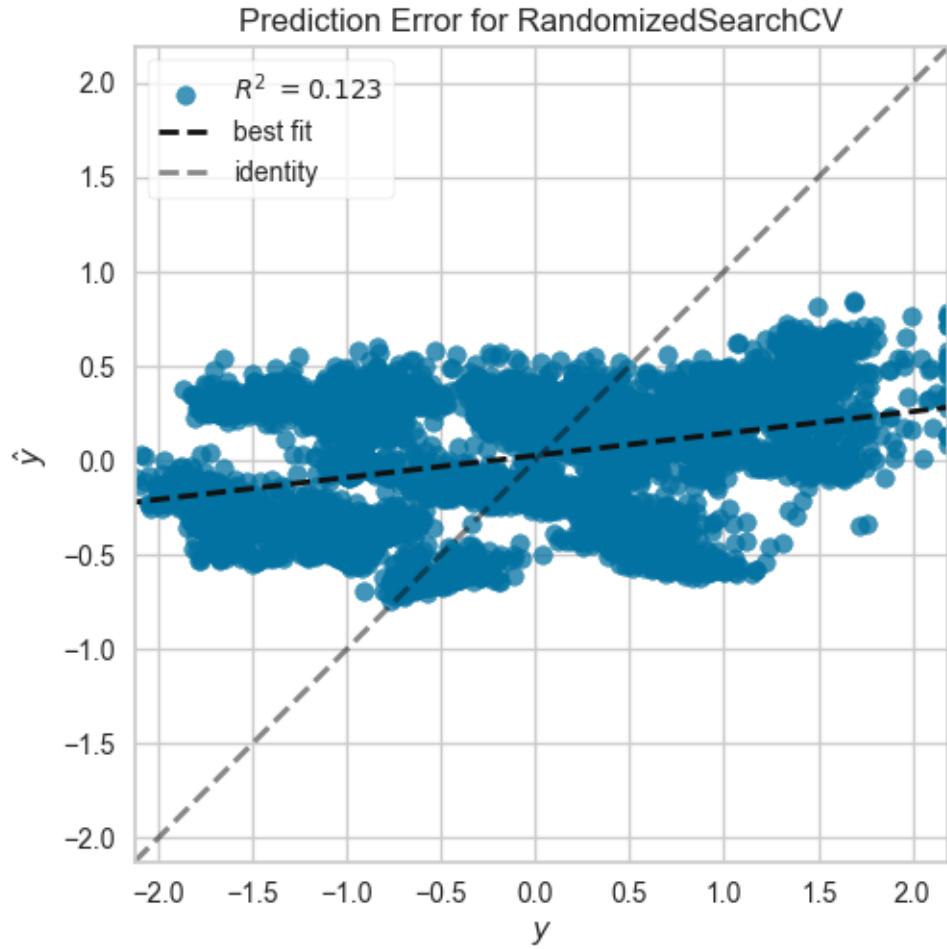
Fitting 5 folds for each of 10 candidates, totalling 50 fits  
[CV] END C=10, average=False, epsilon=0.75025, loss=epsilon\_insensitive,  
shuffle=True; total time= 0.0s



```
[CV] END C=0.001, average=True, epsilon=0.5005,  
loss=squared_epsilon_insensitive, shuffle=False; total time= 0.0s  
[CV] END C=0.001, average=True, epsilon=0.5005,  
loss=squared_epsilon_insensitive, shuffle=False; total time= 0.0s  
[CV] END C=0.001, average=True, epsilon=0.5005,  
loss=squared_epsilon_insensitive, shuffle=False; total time= 0.0s  
[CV] END C=0.001, average=True, epsilon=0.5005,  
loss=squared_epsilon_insensitive, shuffle=False; total time= 0.0s  
[CV] END C=0.001, average=True, epsilon=0.5005,  
loss=squared_epsilon_insensitive, shuffle=False; total time= 0.0s  
[CV] END C=0.01, average=False, epsilon=1.0, loss=epsilon_insensitive,  
shuffle=False; total time= 0.0s  
[CV] END C=0.01, average=False, epsilon=1.0, loss=epsilon_insensitive,  
shuffle=False; total time= 0.0s  
[CV] END C=0.01, average=False, epsilon=1.0, loss=epsilon_insensitive,  
shuffle=False; total time= 0.0s  
[CV] END C=0.01, average=False, epsilon=1.0, loss=epsilon_insensitive,  
shuffle=False; total time= 0.0s  
[CV] END C=0.0001, average=False, epsilon=1.0, loss=epsilon_insensitive,  
shuffle=False; total time= 0.0s  
[CV] END C=0.0001, average=False, epsilon=1.0, loss=epsilon_insensitive,  
shuffle=False; total time= 0.0s  
[CV] END C=0.0001, average=False, epsilon=1.0, loss=epsilon_insensitive,  
shuffle=False; total time= 0.0s  
[CV] END C=0.1, average=False, epsilon=0.75025,  
loss=squared_epsilon_insensitive, shuffle=False; total time= 0.0s  
[CV] END C=0.1, average=False, epsilon=0.75025,  
loss=squared_epsilon_insensitive, shuffle=False; total time= 0.0s  
[CV] END C=0.1, average=False, epsilon=0.75025,  
loss=squared_epsilon_insensitive, shuffle=False; total time= 0.0s  
[CV] END C=0.1, average=False, epsilon=0.75025,  
loss=squared_epsilon_insensitive, shuffle=False; total time= 0.0s  
[CV] END C=10, average=True, epsilon=1.0, loss=epsilon_insensitive,  
shuffle=True; total time= 0.0s  
[CV] END C=10, average=True, epsilon=1.0, loss=epsilon_insensitive,  
shuffle=True; total time= 0.0s  
[CV] END C=10, average=True, epsilon=1.0, loss=epsilon_insensitive,  
shuffle=True; total time= 0.0s
```

```
[CV] END C=10, average=True, epsilon=1.0, loss=epsilon_insensitive,  
shuffle=True; total time= 0.0s  
Mean Absolute Percentage Error(MAPE): 2.0718371806563853  
Root Mean Squared Error(RMSE): 0.9436517054856786  
R2 Score: 0.13010649055828494
```





```
[120]: grid_pa.best_params_
```

```
[120]: {'shuffle': False,
      'loss': 'squared_epsilon_insensitive',
      'epsilon': 0.25075,
      'average': True,
      'C': 0.1}
```

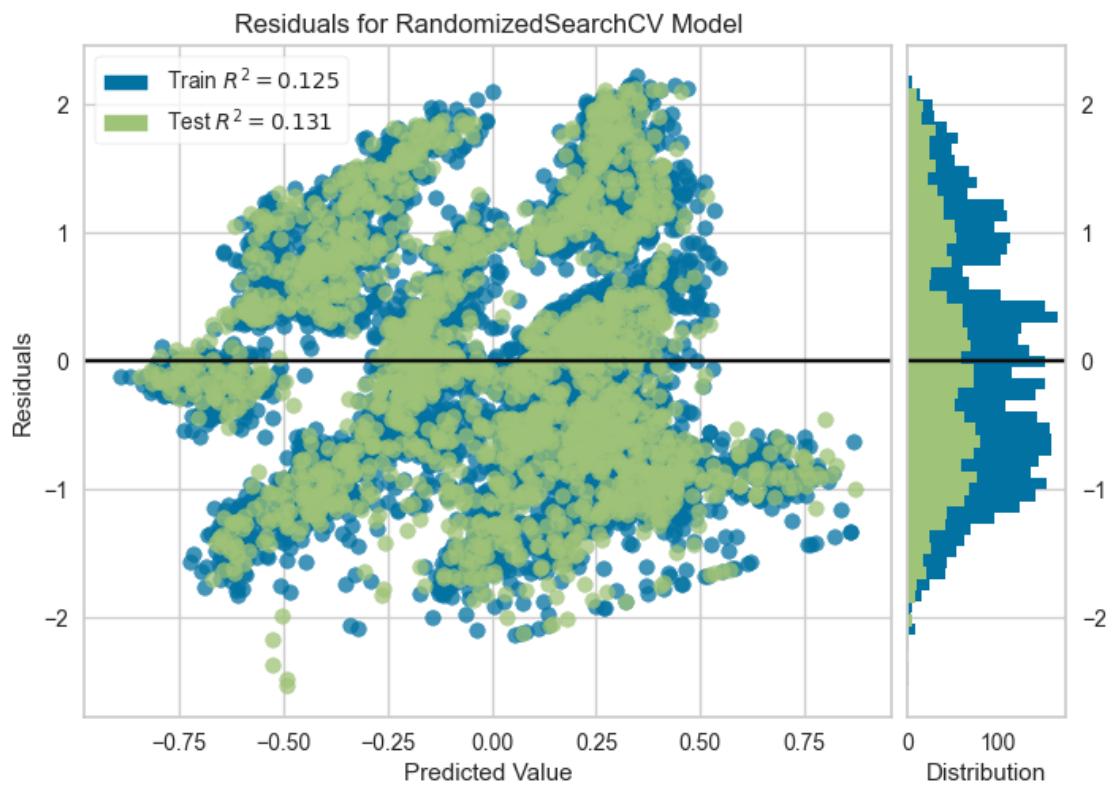
```
[121]: param_grid = {
    'eps': [0.0001, 0.001, 0.01, 0.1, 1],
    'positive': [True, False],
    'selection': ['cyclic', 'random']
}
```

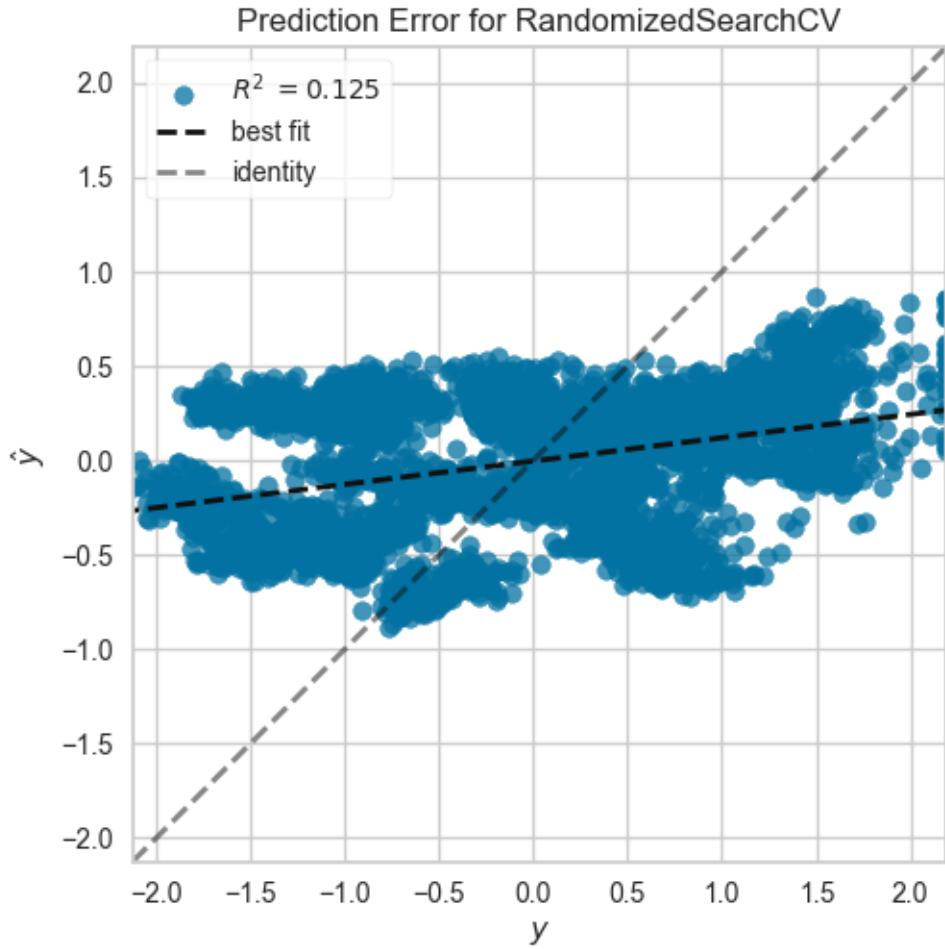
```
grid_lasso = RandomizedSearchCV(LassoCV(), param_grid, verbose=2, cv=5)
train_and_evaluate_model(grid_lasso)
```

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



```
[CV] END ...eps=0.001, positive=True, selection=cyclic; total time= 0.0s
[CV] END ...eps=0.001, positive=True, selection=cyclic; total time= 0.0s
Mean Absolute Percentage Error(MAPE): 1.8794575635430164
Root Mean Squared Error(RMSE): 0.9432143199528643
R2 Score: 0.13091270042859826
```





```
[122]: grid_lasso.best_params_
```

```
[122]: {'selection': 'random', 'positive': False, 'eps': 0.01}
```

```
[123]: param_grid = {
    'alphas': [(0.1, 1.0, 10.0),(0.01,0.1,1),(0.001,0.01,0.1)],
    'gcv_mode': ['auto', 'svd', 'eigen'],
    'store_cv_values': [True, False],
    'alpha_per_target': [True, False]
}
```

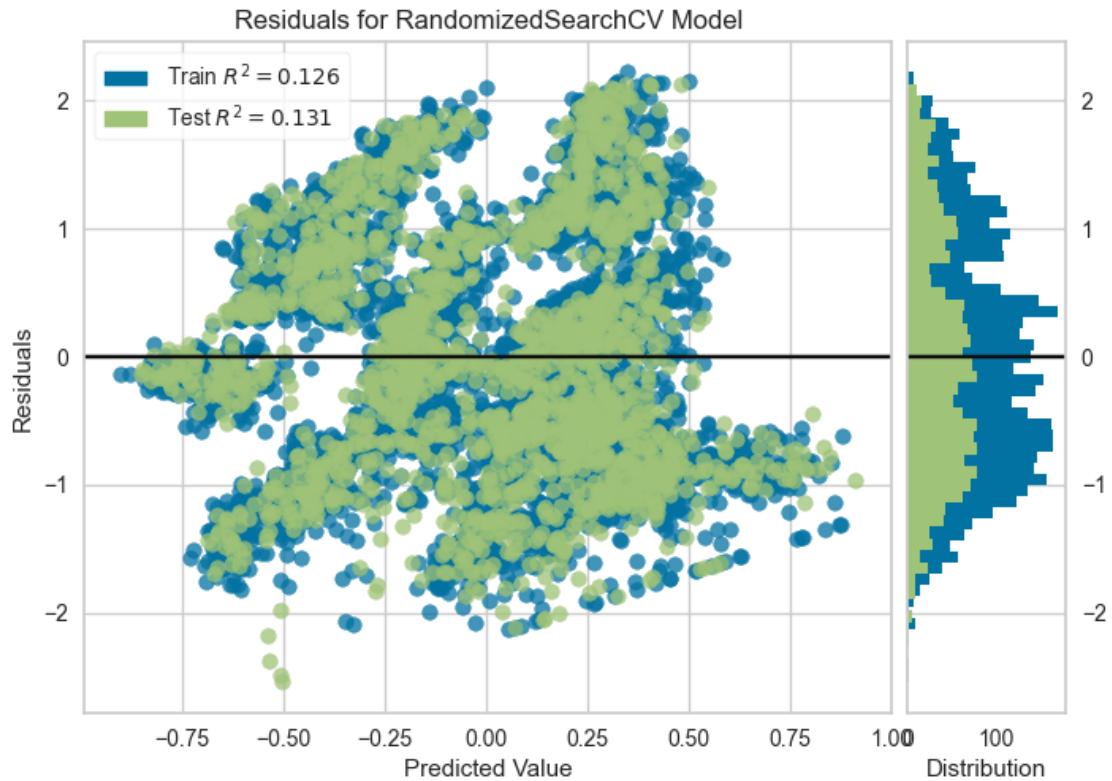
```
grid_ridge = RandomizedSearchCV(RidgeCV(), param_grid, verbose=2, cv=5)
train_and_evaluate_model(grid_ridge)
```

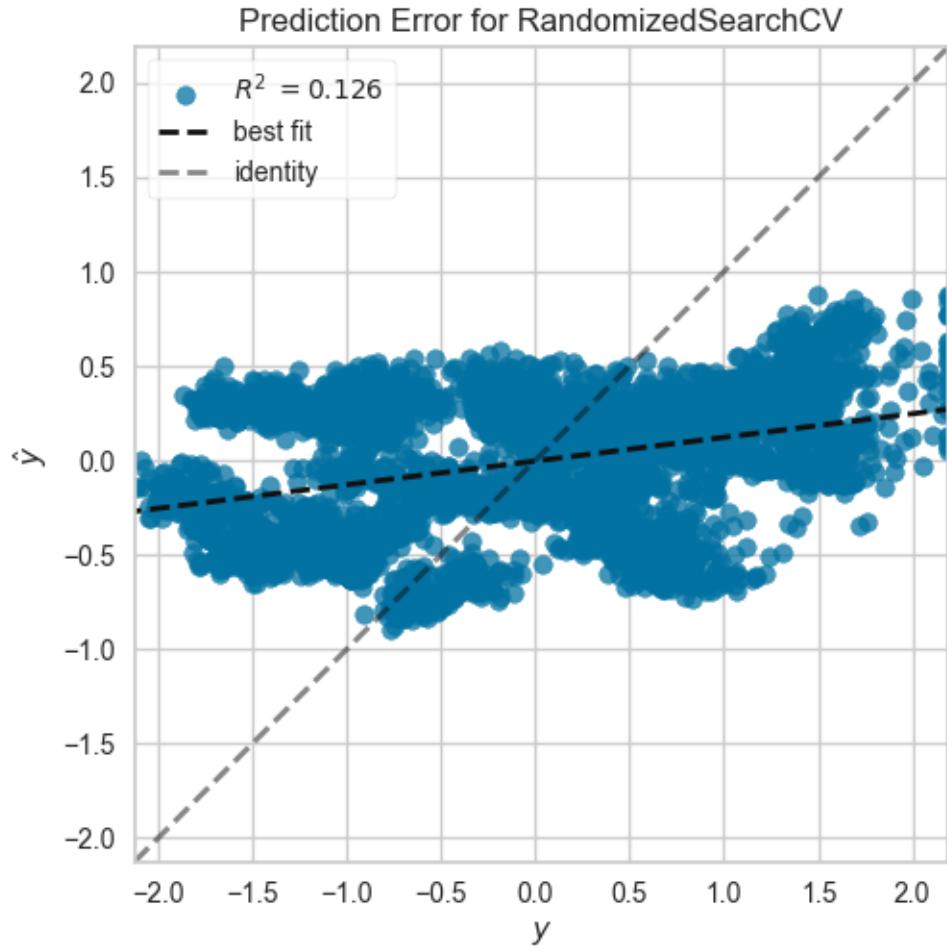
Fitting 5 folds for each of 10 candidates, totalling 50 fits  
[CV] END alpha\_per\_target=False, alphas=(0.001, 0.01, 0.1), gcv\_mode=eigen,  
store\_cv\_values=False; total time= 10.4s  
[CV] END alpha\_per\_target=False, alphas=(0.001, 0.01, 0.1), gcv\_mode=eigen,





```
store_cv_values=False; total time= 0.0s
Mean Absolute Percentage Error(MAPE): 1.8583366137333945
Root Mean Squared Error(RMSE): 0.9429127460413069
R2 Score: 0.1314683581433027
```





```
[124]: grid_ridge.best_params_
```

```
[124]: {'store_cv_values': True,
      'gcv_mode': 'svd',
      'alphas': (0.1, 1.0, 10.0),
      'alpha_per_target': True}
```

```
[125]: param_grid = {'selection': ['cyclic', 'random'],
                  'l1_ratio': [0.1, 0.3, 0.5, 0.8, 1],
                  'positive': [True, False]}
```

```
grid_elasticnet = RandomizedSearchCV(ElasticNetCV(), param_grid, verbose=2, cv=5)
train_and_evaluate_model(grid_elasticnet)
```

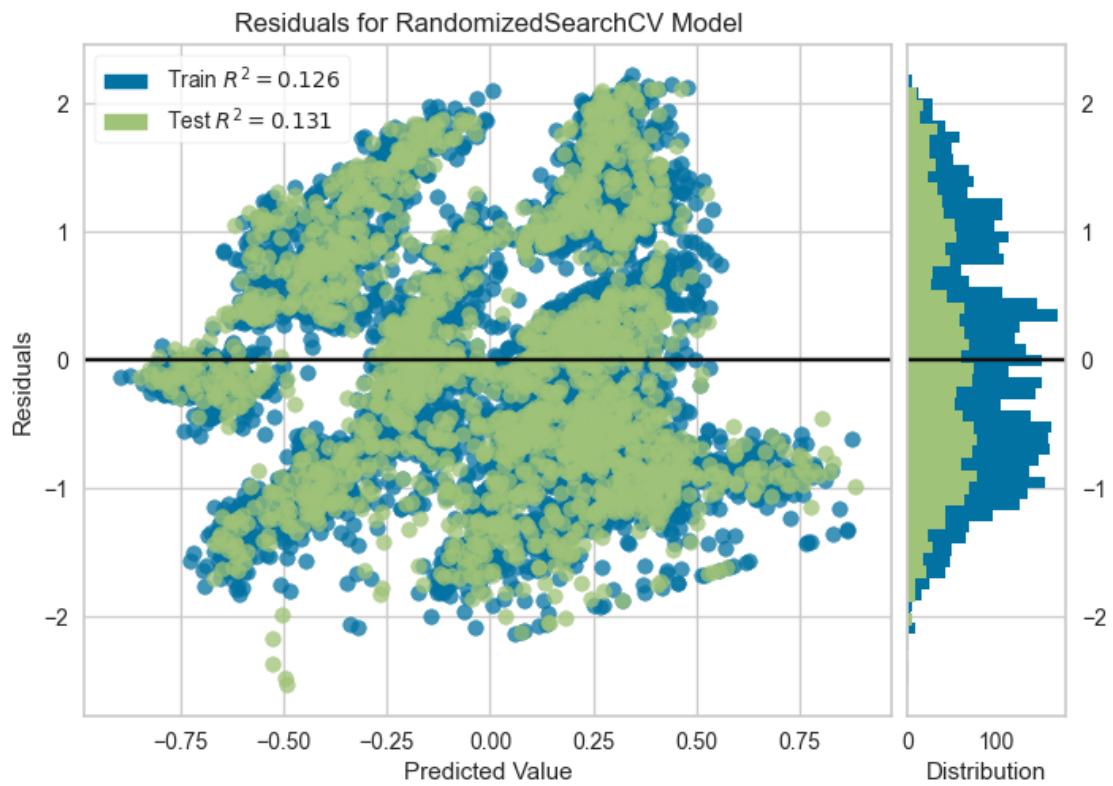
Fitting 5 folds for each of 10 candidates, totalling 50 fits

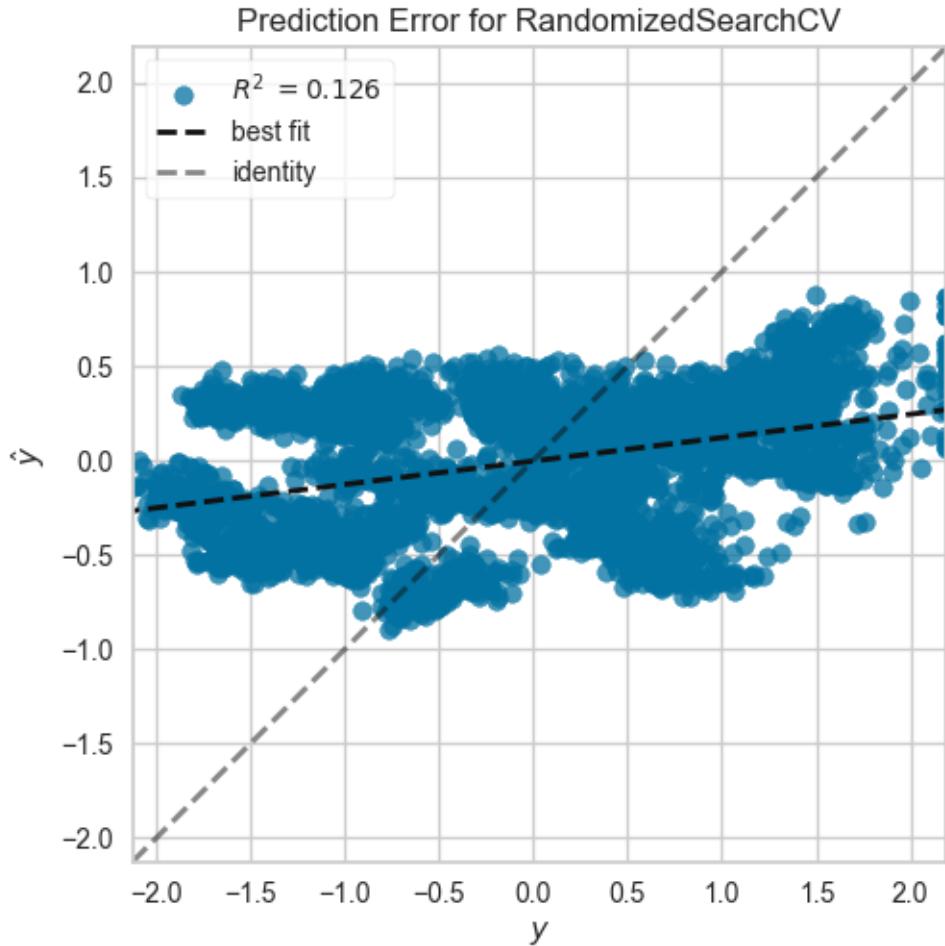
```
[CV] END ...l1_ratio=0.1, positive=True, selection=cyclic; total time= 0.0s
[CV] END ...l1_ratio=0.1, positive=True, selection=cyclic; total time= 0.0s
[CV] END ...l1_ratio=0.1, positive=True, selection=cyclic; total time= 0.0s
```



Root Mean Squared Error(RMSE): 0.9431409062703996

R2 Score: 0.13104798339598622





```
[126]: grid_elasticnet.best_params_
```

```
[126]: {'selection': 'cyclic', 'positive': False, 'l1_ratio': 0.3}
```

```
[127]: param_grid = {'n_estimators': np.arange(100,1001,300),
                  'learning_rate': [0.1,0.4,0.6,0.8,1],
                  'loss': ['linear', 'square', 'exponential']}
```

```
grid_ab = RandomizedSearchCV(AdaBoostRegressor(), param_grid, cv=5, verbose=2)
train_and_evaluate_model(grid_ab)
```

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

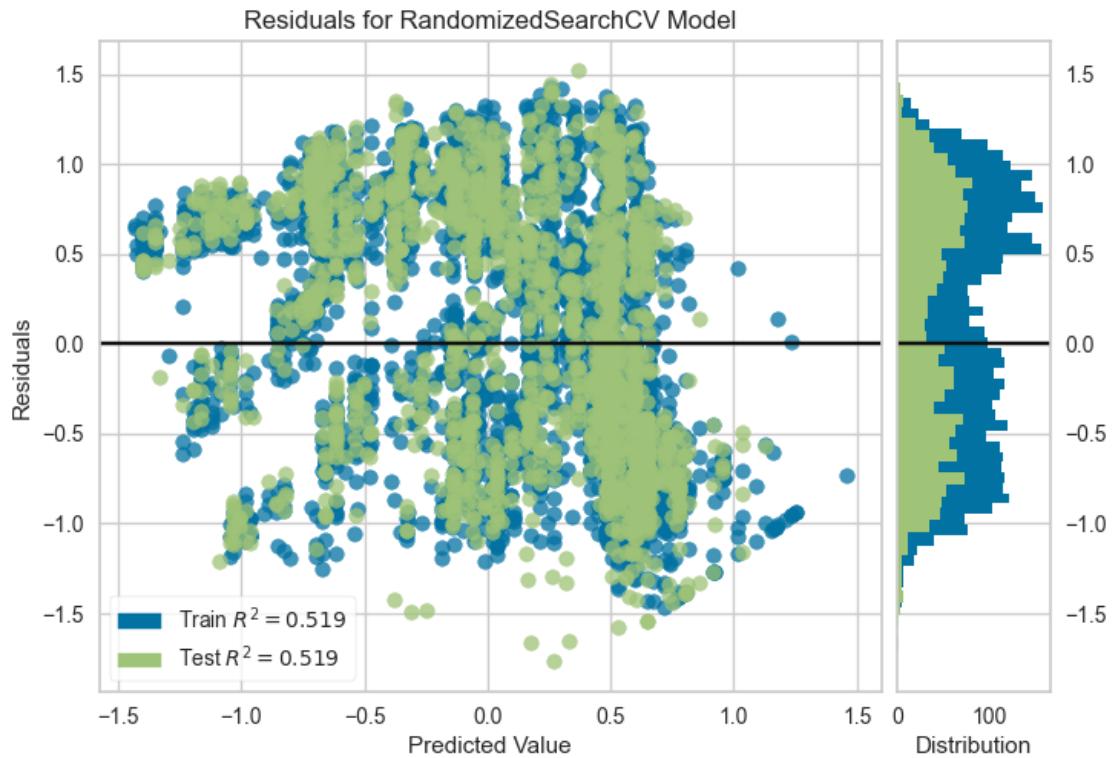
```
[CV] END ...learning_rate=0.4, loss=square, n_estimators=700; total time= 2.5s
[CV] END ...learning_rate=0.4, loss=square, n_estimators=700; total time= 2.2s
[CV] END ...learning_rate=0.4, loss=square, n_estimators=700; total time= 2.2s
[CV] END ...learning_rate=0.4, loss=square, n_estimators=700; total time= 2.2s
[CV] END ...learning_rate=0.4, loss=square, n_estimators=700; total time= 2.1s
[CV] END learning_rate=0.4, loss=exponential, n_estimators=400; total time=
```

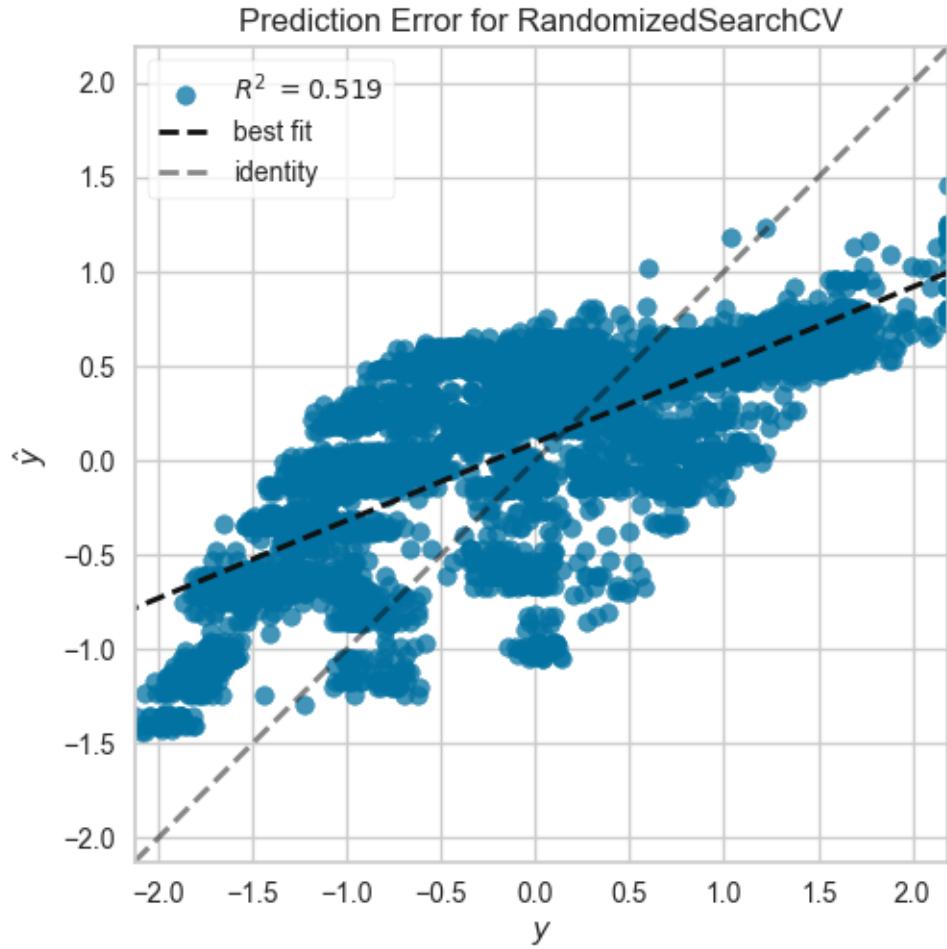


```

[CV] END learning_rate=1, loss=exponential, n_estimators=400; total time= 1.2s
[CV] END ...learning_rate=0.4, loss=square, n_estimators=100; total time= 0.3s
[CV] END learning_rate=0.1, loss=exponential, n_estimators=1000; total time=
4.0s
[CV] END learning_rate=0.1, loss=exponential, n_estimators=1000; total time=
4.0s
[CV] END learning_rate=0.1, loss=exponential, n_estimators=1000; total time=
4.2s
[CV] END learning_rate=0.1, loss=exponential, n_estimators=1000; total time=
4.5s
[CV] END learning_rate=0.1, loss=exponential, n_estimators=1000; total time=
4.3s
Mean Absolute Percentage Error(MAPE): 3.300458874356672
Root Mean Squared Error(RMSE): 0.7017717032881801
R2 Score: 0.5189011292491548

```





```
[128]: grid_ab.best_params_
```

```
[128]: {'n_estimators': 100, 'loss': 'square', 'learning_rate': 1}
```

```
[129]: param_grid = {'alpha_1': [1e-5, 1e-6, 1e-7, 1e-8],
                  'alpha_2': [1e-5, 1e-6, 1e-7, 1e-8],
                  'lambda_1': [1e-5, 1e-6, 1e-7, 1e-8],
                  'lambda_2': [1e-5, 1e-6, 1e-7, 1e-8],
                  'fit_intercept': [True, False],
                  'compute_score': [True, False]}
```

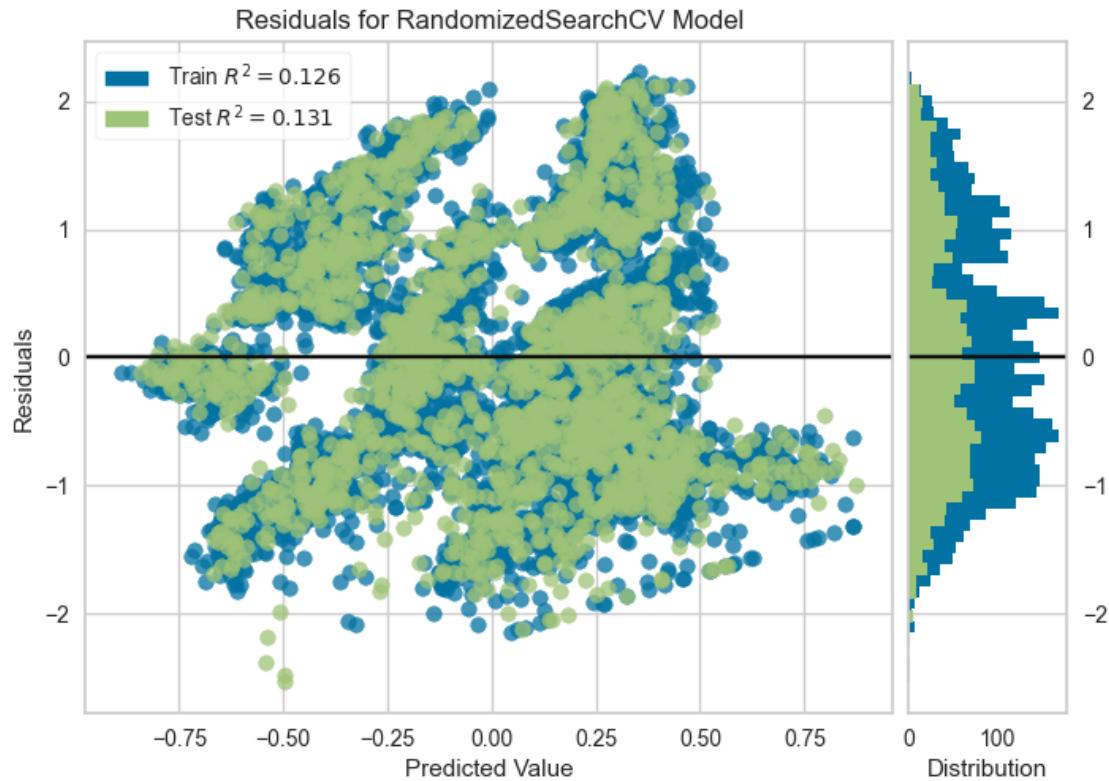
```
grid_ard = RandomizedSearchCV(ARDRegression(), param_grid, verbose=2, cv=5)
train_and_evaluate_model(grid_ard)
```

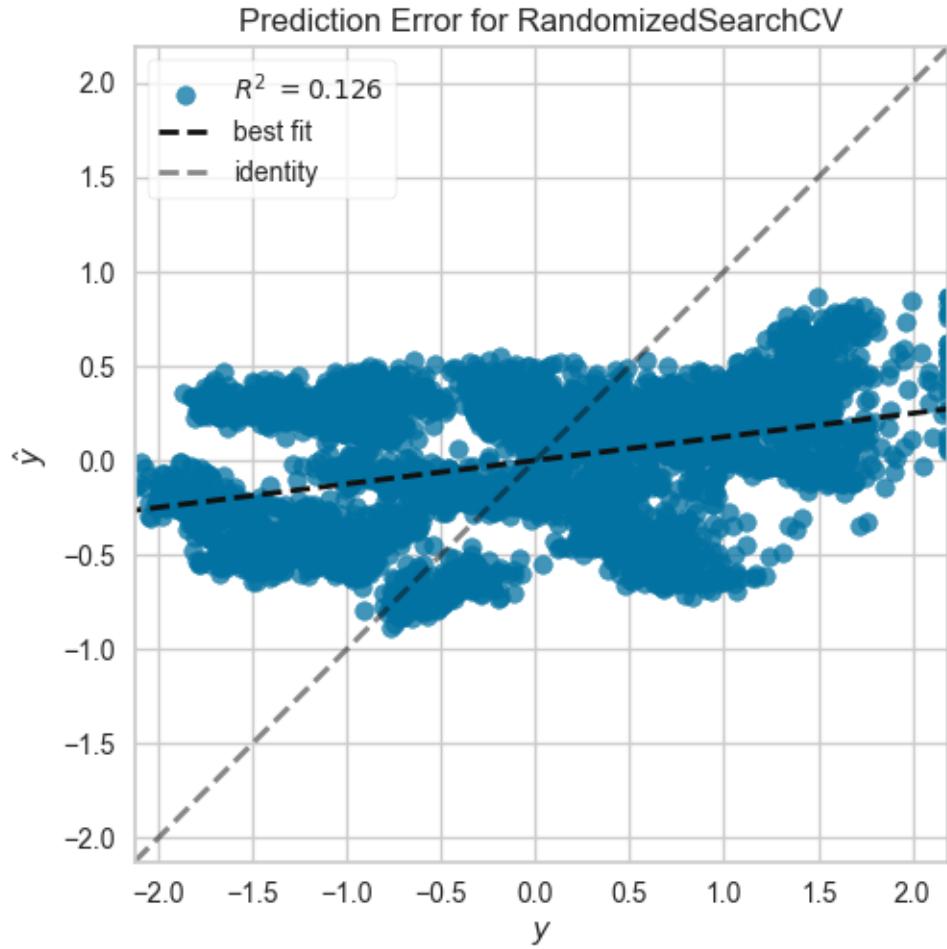
Fitting 5 folds for each of 10 candidates, totalling 50 fits  
[CV] END alpha\_1=1e-06, alpha\_2=1e-06, compute\_score=False, fit\_intercept=False,  
lambda\_1=1e-07, lambda\_2=1e-08; total time= 0.0s  
[CV] END alpha\_1=1e-06, alpha\_2=1e-06, compute\_score=False, fit\_intercept=False,





```
lambda_1=1e-06, lambda_2=1e-06; total time= 0.0s
Mean Absolute Percentage Error(MAPE): 1.8923249460952503
Root Mean Squared Error(RMSE): 0.9431019784663416
R2 Score: 0.13111971328391103
```





```
[130]: grid_ard.best_params_
```

```
[130]: {'lambda_2': 1e-08,
         'lambda_1': 1e-07,
         'fit_intercept': False,
         'compute_score': False,
         'alpha_2': 1e-06,
         'alpha_1': 1e-06}
```

```
[131]: param_grid = {'fit_intercept': [True, False],
                    'positive': [True, False]}
```

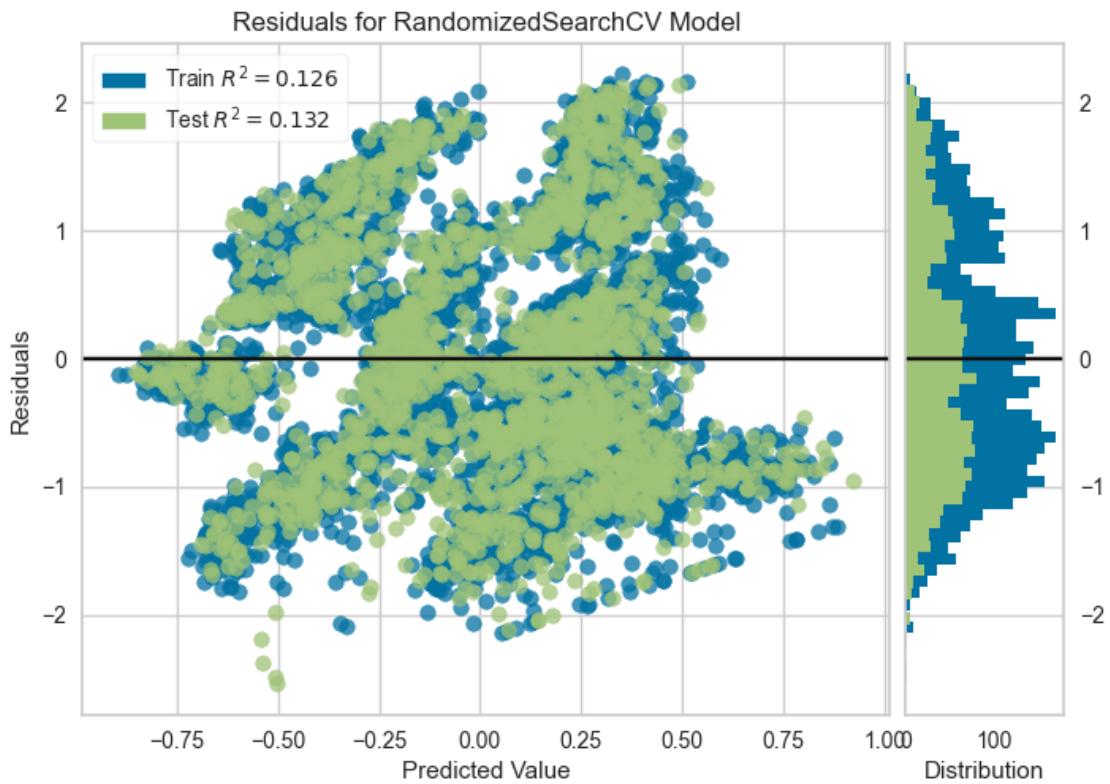
```
grid_lr = RandomizedSearchCV(LinearRegression(), param_grid, verbose=2, cv=5)
train_and_evaluate_model(grid_lr)
```

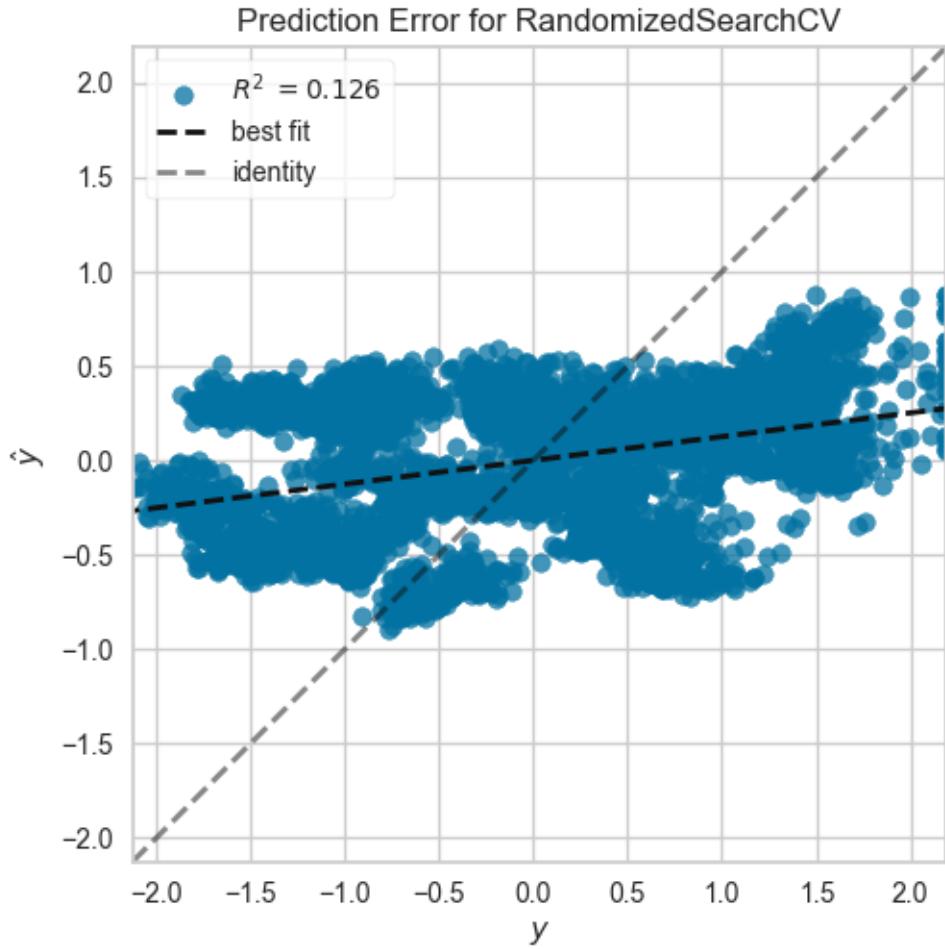
Fitting 5 folds for each of 4 candidates, totalling 20 fits  
[CV] END ...fit\_intercept=True, positive=True; total time= 0.0s  
[CV] END ...fit\_intercept=True, positive=True; total time= 0.0s

```

[CV] END ...fit_intercept=True, positive=True; total time= 0.0s
[CV] END ...fit_intercept=True, positive=True; total time= 0.0s
[CV] END ...fit_intercept=True, positive=True; total time= 0.0s
[CV] END ...fit_intercept=True, positive=False; total time= 0.0s
[CV] END ...fit_intercept=False, positive=True; total time= 0.0s
[CV] END ...fit_intercept=False, positive=True; total time= 0.0s
[CV] END ...fit_intercept=False, positive=True; total time= 0.0s
[CV] END ...fit_intercept=False, positive=False; total time= 0.0s
Mean Absolute Percentage Error(MAPE): 1.8548642518330756
Root Mean Squared Error(RMSE): 0.9427826687072169
R2 Score: 0.13170797413816493

```





```
[132]: grid_lr.best_params_
```

```
[132]: {'positive': False, 'fit_intercept': False}
```

## 0.9 Optimized Models Performance Comparison

```
[133]: model_perfs = pd.DataFrame({'Model': models, 'MAPE': mape_scores, 'RMSE': rmse_scores, 'R2': r2_scores}).sort_values('R2', ascending=False)
model_perfs
```

	Model	MAPE	RMSE	\
34	RandomizedSearchCV(cv=5, \n	e...	0.439785	0.116884
25	<catboost.core.CatBoostRegressor object at 0x0...	0.534145	0.122909	
21	XGBRegressor(base_score=None, booster=None, ca...	0.448424	0.127983	
29	RandomizedSearchCV(cv=RepeatedKFold(n_repeats=...	0.353994	0.128688	
24	LGBMRegressor()	0.483647	0.128976	
18	HistGradientBoostingRegressor()	0.510873	0.129433	

```

26 VotingRegressor(estimators=[('CAT', \n          ... 0.483340 0.131280
36 RandomizedSearchCV(estimator=LGBMRegressor(), \... 0.812379 0.166020
33 RandomizedSearchCV(cv=6, estimator=GradientBoo... 0.538507 0.168305
16 (ExtraTreeRegressor(random_state=1040630546), ... 0.596846 0.178494
37 RandomizedSearchCV(cv=5, estimator=ExtraTreesR... 0.642837 0.179146
35 RandomizedSearchCV(cv=5, estimator=BaggingRegr... 0.625728 0.179327
14 (DecisionTreeRegressor(max_features=1.0, rando... 0.631636 0.180889
15 (DecisionTreeRegressor(random_state=1012004194... 0.628884 0.181817
27 RandomizedSearchCV(cv=5, estimator=RandomFores... 0.670434 0.216686
32 RandomizedSearchCV(cv=RepeatedKFold(n_repeats=... 0.679039 0.220655
13                               DecisionTreeRegressor() 0.688617 0.229284
19 ([DecisionTreeRegressor(criterion='friedman_ms... 1.525258 0.302092
22 XGBRFRegressor(base_score=None, booster=None, ... 1.495870 0.395600
28 RandomizedSearchCV(cv=RepeatedKFold(n_repeats=... 1.827042 0.678111
20                               MLPRegressor() 2.302538 0.680757
46 RandomizedSearchCV(cv=5, estimator=AdaBoostReg... 3.300459 0.701772
17 (DecisionTreeRegressor(max_depth=3, random_sta... 2.575329 0.725043
10                               KNeighborsRegressor() 1.832426 0.725657
23                               NuSVR() 2.767677 0.795309
30 RandomizedSearchCV(cv=5, estimator=SVR(), \n    ... 2.015902 0.806134
11                               SVR() 1.778644 0.815938
48 RandomizedSearchCV(cv=5, estimator=LinearRegres... 1.854864 0.942783
0                               LinearRegression() 1.858860 0.942806
40 RandomizedSearchCV(cv=5, estimator=HuberRegres... 1.858399 0.942907
2                               RidgeCV() 1.858337 0.942913
44 RandomizedSearchCV(cv=5, estimator=RidgeCV(), \... 1.858337 0.942913
4                               SGDRegressor() 1.824808 0.942968
38 RandomizedSearchCV(cv=5, \n          ... 1.864210 0.942998
1                               LassoCV() 1.872174 0.943096
47 RandomizedSearchCV(cv=5, estimator=ARDRegressi... 1.892325 0.943102
3                               ElasticNetCV() 1.869417 0.943112
6                               ARDRegression() 1.896284 0.943125
45 RandomizedSearchCV(cv=5, estimator=ElasticNetC... 1.866485 0.943141
43 RandomizedSearchCV(cv=5, estimator=LassoCV(), \... 1.879458 0.943214
42 RandomizedSearchCV(cv=5, estimator=PassiveAggr... 2.071837 0.943652
5                               HuberRegressor() 2.047510 0.945681
31 RandomizedSearchCV(cv=RepeatedKFold(n_repeats=... 1.802767 0.945701
12                               LinearSVR() 2.058692 0.953704
39 RandomizedSearchCV(cv=5, estimator=SGDRegresso... 2.256280 0.964216
8                               TweedieRegressor() 1.300327 0.964521
41 RandomizedSearchCV(cv=5, estimator=TweedieRegr... 1.053728 0.994684
9                               PassiveAggressiveRegressor() 5.203633 1.123522
7                               RANSACRegressor() 4.191927 1.257529

```

R2

```

34 0.986654
25 0.985243

```

21	0.983999
29	0.983822
24	0.983750
18	0.983634
26	0.983164
36	0.973074
33	0.972328
16	0.968876
37	0.968649
35	0.968585
14	0.968036
15	0.967707
27	0.954133
32	0.952437
13	0.948644
19	0.910850
22	0.847118
28	0.550795
20	0.547282
46	0.518901
17	0.486465
10	0.485594
23	0.382105
30	0.365171
11	0.349636
48	0.131708
0	0.131666
40	0.131479
2	0.131468
44	0.131468
4	0.131367
38	0.131312
1	0.131130
47	0.131120
3	0.131102
6	0.131078
45	0.131048
43	0.130913
42	0.130106
5	0.126361
31	0.126324
12	0.111474
39	0.091779
8	0.091205
41	0.033475
9	-0.233122
7	-0.544823

Hyperparameter-tuned Cat Boost Regressor is the best performing model among all optimized models, achieving an outstanding r2 score of nearly 98.7%.

## 0.10 Saving the best model for future use

```
[141]: # Cross validating the performance of the Cat Boost Regressor model before ↴ saving it
avg_cv_score = np.
    ↴ mean(cross_val_score(grid_cat,X_test,y_test,scoring='r2',cv=RepeatedKFold(n_splits=4,n_repe
print("Average cross validation R2 score:",avg_cv_score)
```

[Parallel(n\_jobs=1)]: Using backend SequentialBackend with 1 concurrent workers.

Fitting 5 folds for each of 10 candidates, totalling 50 fits  
[CV 1/5] END learning\_rate=0.4, n\_estimators=700;, score=0.967 total time= 1.6s  
[CV 2/5] END learning\_rate=0.4, n\_estimators=700;, score=0.962 total time= 1.2s  
[CV 3/5] END learning\_rate=0.4, n\_estimators=700;, score=0.967 total time= 1.3s  
[CV 4/5] END learning\_rate=0.4, n\_estimators=700;, score=0.957 total time= 1.2s  
[CV 5/5] END learning\_rate=0.4, n\_estimators=700;, score=0.962 total time= 1.3s  
[CV 1/5] END learning\_rate=0.4, n\_estimators=400;, score=0.967 total time= 1.1s  
[CV 2/5] END learning\_rate=0.4, n\_estimators=400;, score=0.962 total time= 1.0s  
[CV 3/5] END learning\_rate=0.4, n\_estimators=400;, score=0.967 total time= 0.9s  
[CV 4/5] END learning\_rate=0.4, n\_estimators=400;, score=0.957 total time= 0.9s  
[CV 5/5] END learning\_rate=0.4, n\_estimators=400;, score=0.961 total time= 0.9s  
[CV 1/5] END learning\_rate=0.2, n\_estimators=400;, score=0.967 total time= 1.1s  
[CV 2/5] END learning\_rate=0.2, n\_estimators=400;, score=0.968 total time= 0.9s  
[CV 3/5] END learning\_rate=0.2, n\_estimators=400;, score=0.973 total time= 0.9s  
[CV 4/5] END learning\_rate=0.2, n\_estimators=400;, score=0.967 total time= 0.9s  
[CV 5/5] END learning\_rate=0.2, n\_estimators=400;, score=0.964 total time= 0.9s  
[CV 1/5] END learning\_rate=0.7, n\_estimators=1000;, score=0.963 total time= 1.7s  
[CV 2/5] END learning\_rate=0.7, n\_estimators=1000;, score=0.960 total time= 1.7s

```

[CV 3/5] END learning_rate=0.7, n_estimators=1000;, score=0.963 total time=
1.8s
[CV 4/5] END learning_rate=0.7, n_estimators=1000;, score=0.965 total time=
2.2s
[CV 5/5] END learning_rate=0.7, n_estimators=1000;, score=0.948 total time=
1.5s
[CV 1/5] END learning_rate=0.5, n_estimators=100;, score=0.961 total time=
0.6s
[CV 2/5] END learning_rate=0.5, n_estimators=100;, score=0.954 total time=
0.7s
[CV 3/5] END learning_rate=0.5, n_estimators=100;, score=0.966 total time=
0.7s
[CV 4/5] END learning_rate=0.5, n_estimators=100;, score=0.957 total time=
0.8s
[CV 5/5] END learning_rate=0.5, n_estimators=100;, score=0.958 total time=
0.5s
[CV 1/5] END .learning_rate=1, n_estimators=100;, score=0.941 total time= 0.6s
[CV 2/5] END .learning_rate=1, n_estimators=100;, score=0.933 total time= 0.5s
[CV 3/5] END .learning_rate=1, n_estimators=100;, score=0.940 total time= 0.4s
[CV 4/5] END .learning_rate=1, n_estimators=100;, score=0.941 total time= 0.8s
[CV 5/5] END .learning_rate=1, n_estimators=100;, score=0.922 total time= 0.6s
[CV 1/5] END learning_rate=0.4, n_estimators=100;, score=0.962 total time=
0.7s
[CV 2/5] END learning_rate=0.4, n_estimators=100;, score=0.955 total time=
0.7s
[CV 3/5] END learning_rate=0.4, n_estimators=100;, score=0.966 total time=
0.7s
[CV 4/5] END learning_rate=0.4, n_estimators=100;, score=0.953 total time=
0.5s
[CV 5/5] END learning_rate=0.4, n_estimators=100;, score=0.956 total time=
0.6s
[CV 1/5] END learning_rate=0.2, n_estimators=1000;, score=0.968 total time=
1.8s
[CV 2/5] END learning_rate=0.2, n_estimators=1000;, score=0.969 total time=
1.8s
[CV 3/5] END learning_rate=0.2, n_estimators=1000;, score=0.973 total time=
1.6s
[CV 4/5] END learning_rate=0.2, n_estimators=1000;, score=0.968 total time=
1.8s
[CV 5/5] END learning_rate=0.2, n_estimators=1000;, score=0.964 total time=
1.9s
[CV 1/5] END .learning_rate=1, n_estimators=400;, score=0.942 total time= 1.1s
[CV 2/5] END .learning_rate=1, n_estimators=400;, score=0.933 total time= 1.1s
[CV 3/5] END .learning_rate=1, n_estimators=400;, score=0.941 total time= 1.1s
[CV 4/5] END .learning_rate=1, n_estimators=400;, score=0.942 total time= 1.2s
[CV 5/5] END .learning_rate=1, n_estimators=400;, score=0.923 total time= 1.1s
[CV 1/5] END .learning_rate=1, n_estimators=700;, score=0.942 total time= 1.4s
[CV 2/5] END .learning_rate=1, n_estimators=700;, score=0.934 total time= 1.5s

```

```

[CV 3/5] END .learning_rate=1, n_estimators=700;, score=0.941 total time= 1.6s
[CV 4/5] END .learning_rate=1, n_estimators=700;, score=0.942 total time= 1.4s
[CV 5/5] END .learning_rate=1, n_estimators=700;, score=0.923 total time= 1.5s
Fitting 5 folds for each of 10 candidates, totalling 50 fits
[CV 1/5] END learning_rate=0.5, n_estimators=100;, score=0.949 total time=
0.6s
[CV 2/5] END learning_rate=0.5, n_estimators=100;, score=0.966 total time=
0.8s
[CV 3/5] END learning_rate=0.5, n_estimators=100;, score=0.969 total time=
1.0s
[CV 4/5] END learning_rate=0.5, n_estimators=100;, score=0.951 total time=
0.8s
[CV 5/5] END learning_rate=0.5, n_estimators=100;, score=0.964 total time=
0.8s
[CV 1/5] END learning_rate=0.2, n_estimators=800;, score=0.958 total time=
1.5s
[CV 2/5] END learning_rate=0.2, n_estimators=800;, score=0.972 total time=
1.5s
[CV 3/5] END learning_rate=0.2, n_estimators=800;, score=0.973 total time=
1.6s
[CV 4/5] END learning_rate=0.2, n_estimators=800;, score=0.966 total time=
1.7s
[CV 5/5] END learning_rate=0.2, n_estimators=800;, score=0.972 total time=
1.4s
[CV 1/5] END learning_rate=0.4, n_estimators=400;, score=0.953 total time=
1.6s
[CV 2/5] END learning_rate=0.4, n_estimators=400;, score=0.961 total time=
1.4s
[CV 3/5] END learning_rate=0.4, n_estimators=400;, score=0.966 total time=
1.2s
[CV 4/5] END learning_rate=0.4, n_estimators=400;, score=0.952 total time=
1.1s
[CV 5/5] END learning_rate=0.4, n_estimators=400;, score=0.971 total time=
1.2s
[CV 1/5] END learning_rate=0.7, n_estimators=400;, score=0.941 total time=
1.2s
[CV 2/5] END learning_rate=0.7, n_estimators=400;, score=0.958 total time=
1.3s
[CV 3/5] END learning_rate=0.7, n_estimators=400;, score=0.956 total time=
1.1s
[CV 4/5] END learning_rate=0.7, n_estimators=400;, score=0.942 total time=
1.2s
[CV 5/5] END learning_rate=0.7, n_estimators=400;, score=0.952 total time=
1.3s
[CV 1/5] END .learning_rate=1, n_estimators=100;, score=0.952 total time= 1.0s
[CV 2/5] END .learning_rate=1, n_estimators=100;, score=0.940 total time= 0.8s
[CV 3/5] END .learning_rate=1, n_estimators=100;, score=0.935 total time= 0.8s
[CV 4/5] END .learning_rate=1, n_estimators=100;, score=0.937 total time= 0.7s

```

```
[CV 5/5] END .learning_rate=1, n_estimators=100;, score=0.943 total time= 0.7s
[CV 1/5] END learning_rate=0.4, n_estimators=1000;, score=0.952 total time=
1.9s
[CV 2/5] END learning_rate=0.4, n_estimators=1000;, score=0.961 total time=
1.8s
[CV 3/5] END learning_rate=0.4, n_estimators=1000;, score=0.967 total time=
1.9s
[CV 4/5] END learning_rate=0.4, n_estimators=1000;, score=0.952 total time=
2.2s
[CV 5/5] END learning_rate=0.4, n_estimators=1000;, score=0.971 total time=
2.0s
[CV 1/5] END learning_rate=0.7, n_estimators=700;, score=0.942 total time=
1.4s
[CV 2/5] END learning_rate=0.7, n_estimators=700;, score=0.958 total time=
1.4s
[CV 3/5] END learning_rate=0.7, n_estimators=700;, score=0.956 total time=
1.3s
[CV 4/5] END learning_rate=0.7, n_estimators=700;, score=0.942 total time=
1.4s
[CV 5/5] END learning_rate=0.7, n_estimators=700;, score=0.952 total time=
1.7s
[CV 1/5] END learning_rate=0.5, n_estimators=1000;, score=0.952 total time=
2.1s
[CV 2/5] END learning_rate=0.5, n_estimators=1000;, score=0.968 total time=
2.0s
[CV 3/5] END learning_rate=0.5, n_estimators=1000;, score=0.972 total time=
1.8s
[CV 4/5] END learning_rate=0.5, n_estimators=1000;, score=0.956 total time=
2.2s
[CV 5/5] END learning_rate=0.5, n_estimators=1000;, score=0.967 total time=
2.1s
[CV 1/5] END learning_rate=0.5, n_estimators=700;, score=0.952 total time=
1.5s
[CV 2/5] END learning_rate=0.5, n_estimators=700;, score=0.968 total time=
1.4s
[CV 3/5] END learning_rate=0.5, n_estimators=700;, score=0.972 total time=
1.8s
[CV 4/5] END learning_rate=0.5, n_estimators=700;, score=0.956 total time=
1.5s
[CV 5/5] END learning_rate=0.5, n_estimators=700;, score=0.967 total time=
1.5s
[CV 1/5] END learning_rate=0.5, n_estimators=800;, score=0.952 total time=
1.6s
[CV 2/5] END learning_rate=0.5, n_estimators=800;, score=0.968 total time=
1.8s
[CV 3/5] END learning_rate=0.5, n_estimators=800;, score=0.972 total time=
1.4s
[CV 4/5] END learning_rate=0.5, n_estimators=800;, score=0.956 total time=
```

```

1.5s
[CV 5/5] END learning_rate=0.5, n_estimators=800;, score=0.967 total time=
1.6s
Fitting 5 folds for each of 10 candidates, totalling 50 fits
[CV 1/5] END learning_rate=0.2, n_estimators=100;, score=0.944 total time=
0.7s
[CV 2/5] END learning_rate=0.2, n_estimators=100;, score=0.952 total time=
0.7s
[CV 3/5] END learning_rate=0.2, n_estimators=100;, score=0.966 total time=
0.8s
[CV 4/5] END learning_rate=0.2, n_estimators=100;, score=0.952 total time=
0.8s
[CV 5/5] END learning_rate=0.2, n_estimators=100;, score=0.947 total time=
1.0s
[CV 1/5] END .learning_rate=1, n_estimators=100;, score=0.920 total time= 0.8s
[CV 2/5] END .learning_rate=1, n_estimators=100;, score=0.948 total time= 0.7s
[CV 3/5] END .learning_rate=1, n_estimators=100;, score=0.959 total time= 0.7s
[CV 4/5] END .learning_rate=1, n_estimators=100;, score=0.897 total time= 0.6s
[CV 5/5] END .learning_rate=1, n_estimators=100;, score=0.948 total time= 0.7s
[CV 1/5] END learning_rate=0.2, n_estimators=800;, score=0.966 total time=
1.7s
[CV 2/5] END learning_rate=0.2, n_estimators=800;, score=0.972 total time=
1.5s
[CV 3/5] END learning_rate=0.2, n_estimators=800;, score=0.976 total time=
1.7s
[CV 4/5] END learning_rate=0.2, n_estimators=800;, score=0.970 total time=
1.6s
[CV 5/5] END learning_rate=0.2, n_estimators=800;, score=0.969 total time=
1.6s
[CV 1/5] END learning_rate=0.7, n_estimators=700;, score=0.960 total time=
1.4s
[CV 2/5] END learning_rate=0.7, n_estimators=700;, score=0.963 total time=
1.5s
[CV 3/5] END learning_rate=0.7, n_estimators=700;, score=0.962 total time=
1.5s
[CV 4/5] END learning_rate=0.7, n_estimators=700;, score=0.932 total time=
1.5s
[CV 5/5] END learning_rate=0.7, n_estimators=700;, score=0.961 total time=
1.7s
[CV 1/5] END learning_rate=0.5, n_estimators=100;, score=0.956 total time=
0.8s
[CV 2/5] END learning_rate=0.5, n_estimators=100;, score=0.960 total time=
1.1s
[CV 3/5] END learning_rate=0.5, n_estimators=100;, score=0.975 total time=
0.7s
[CV 4/5] END learning_rate=0.5, n_estimators=100;, score=0.948 total time=
0.7s
[CV 5/5] END learning_rate=0.5, n_estimators=100;, score=0.959 total time=

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0.7s
[CV 1/5] END learning_rate=0.2, n_estimators=1000;, score=0.966 total time=
1.9s
[CV 2/5] END learning_rate=0.2, n_estimators=1000;, score=0.972 total time=
2.1s
[CV 3/5] END learning_rate=0.2, n_estimators=1000;, score=0.977 total time=
1.7s
[CV 4/5] END learning_rate=0.2, n_estimators=1000;, score=0.971 total time=
1.7s
[CV 5/5] END learning_rate=0.2, n_estimators=1000;, score=0.969 total time=
2.1s
[CV 1/5] END learning_rate=0.4, n_estimators=400;, score=0.954 total time=
1.1s
[CV 2/5] END learning_rate=0.4, n_estimators=400;, score=0.967 total time=
1.3s
[CV 3/5] END learning_rate=0.4, n_estimators=400;, score=0.968 total time=
1.1s
[CV 4/5] END learning_rate=0.4, n_estimators=400;, score=0.961 total time=
1.7s
[CV 5/5] END learning_rate=0.4, n_estimators=400;, score=0.958 total time=
1.1s
[CV 1/5] END .learning_rate=1, n_estimators=800;, score=0.921 total time= 1.8s
[CV 2/5] END .learning_rate=1, n_estimators=800;, score=0.947 total time= 1.8s
[CV 3/5] END .learning_rate=1, n_estimators=800;, score=0.960 total time= 1.5s
[CV 4/5] END .learning_rate=1, n_estimators=800;, score=0.901 total time= 1.7s
[CV 5/5] END .learning_rate=1, n_estimators=800;, score=0.949 total time= 1.5s
[CV 1/5] END .learning_rate=1, n_estimators=700;, score=0.921 total time= 1.4s
[CV 2/5] END .learning_rate=1, n_estimators=700;, score=0.947 total time= 1.5s
[CV 3/5] END .learning_rate=1, n_estimators=700;, score=0.960 total time= 2.0s
[CV 4/5] END .learning_rate=1, n_estimators=700;, score=0.901 total time= 1.4s
[CV 5/5] END .learning_rate=1, n_estimators=700;, score=0.949 total time= 1.4s
[CV 1/5] END learning_rate=0.7, n_estimators=1000;, score=0.960 total time=
1.7s
[CV 2/5] END learning_rate=0.7, n_estimators=1000;, score=0.963 total time=
1.7s
[CV 3/5] END learning_rate=0.7, n_estimators=1000;, score=0.962 total time=
1.8s
[CV 4/5] END learning_rate=0.7, n_estimators=1000;, score=0.933 total time=
2.3s
[CV 5/5] END learning_rate=0.7, n_estimators=1000;, score=0.961 total time=
1.7s
Fitting 5 folds for each of 10 candidates, totalling 50 fits
[CV 1/5] END learning_rate=0.5, n_estimators=800;, score=0.969 total time=
1.8s
[CV 2/5] END learning_rate=0.5, n_estimators=800;, score=0.960 total time=
1.5s
[CV 3/5] END learning_rate=0.5, n_estimators=800;, score=0.967 total time=
1.7s

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[CV 4/5] END learning_rate=0.5, n_estimators=800;, score=0.957 total time=
1.5s
[CV 5/5] END learning_rate=0.5, n_estimators=800;, score=0.960 total time=
1.7s
[CV 1/5] END learning_rate=0.7, n_estimators=800;, score=0.972 total time=
1.7s
[CV 2/5] END learning_rate=0.7, n_estimators=800;, score=0.945 total time=
1.6s
[CV 3/5] END learning_rate=0.7, n_estimators=800;, score=0.962 total time=
3.0s
[CV 4/5] END learning_rate=0.7, n_estimators=800;, score=0.950 total time=
2.1s
[CV 5/5] END learning_rate=0.7, n_estimators=800;, score=0.952 total time=
1.6s
[CV 1/5] END .learning_rate=1, n_estimators=800;, score=0.939 total time= 1.6s
[CV 2/5] END .learning_rate=1, n_estimators=800;, score=0.942 total time= 1.5s
[CV 3/5] END .learning_rate=1, n_estimators=800;, score=0.938 total time= 2.2s
[CV 4/5] END .learning_rate=1, n_estimators=800;, score=0.929 total time= 1.7s
[CV 5/5] END .learning_rate=1, n_estimators=800;, score=0.936 total time= 1.6s
[CV 1/5] END .learning_rate=1, n_estimators=700;, score=0.939 total time= 1.5s
[CV 2/5] END .learning_rate=1, n_estimators=700;, score=0.942 total time= 1.4s
[CV 3/5] END .learning_rate=1, n_estimators=700;, score=0.938 total time= 1.5s
[CV 4/5] END .learning_rate=1, n_estimators=700;, score=0.929 total time= 1.4s
[CV 5/5] END .learning_rate=1, n_estimators=700;, score=0.936 total time= 1.5s
[CV 1/5] END learning_rate=0.2, n_estimators=1000;, score=0.975 total time=
1.8s
[CV 2/5] END learning_rate=0.2, n_estimators=1000;, score=0.972 total time=
2.2s
[CV 3/5] END learning_rate=0.2, n_estimators=1000;, score=0.971 total time=
1.9s
[CV 4/5] END learning_rate=0.2, n_estimators=1000;, score=0.966 total time=
2.3s
[CV 5/5] END learning_rate=0.2, n_estimators=1000;, score=0.969 total time=
1.7s
[CV 1/5] END learning_rate=0.2, n_estimators=700;, score=0.976 total time=
1.4s
[CV 2/5] END learning_rate=0.2, n_estimators=700;, score=0.971 total time=
1.4s
[CV 3/5] END learning_rate=0.2, n_estimators=700;, score=0.970 total time=
1.4s
[CV 4/5] END learning_rate=0.2, n_estimators=700;, score=0.965 total time=
1.5s
[CV 5/5] END learning_rate=0.2, n_estimators=700;, score=0.969 total time=
1.5s
[CV 1/5] END learning_rate=1, n_estimators=1000;, score=0.939 total time= 1.8s
[CV 2/5] END learning_rate=1, n_estimators=1000;, score=0.942 total time= 2.3s
[CV 3/5] END learning_rate=1, n_estimators=1000;, score=0.938 total time= 1.7s
[CV 4/5] END learning_rate=1, n_estimators=1000;, score=0.929 total time= 2.3s

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[CV 5/5] END learning_rate=1, n_estimators=1000;, score=0.936 total time= 1.8s
[CV 1/5] END learning_rate=0.5, n_estimators=400;, score=0.969 total time=
1.1s
[CV 2/5] END learning_rate=0.5, n_estimators=400;, score=0.960 total time=
1.1s
[CV 3/5] END learning_rate=0.5, n_estimators=400;, score=0.967 total time=
1.2s
[CV 4/5] END learning_rate=0.5, n_estimators=400;, score=0.956 total time=
1.0s
[CV 5/5] END learning_rate=0.5, n_estimators=400;, score=0.960 total time=
1.1s
[CV 1/5] END learning_rate=0.2, n_estimators=400;, score=0.975 total time=
1.0s
[CV 2/5] END learning_rate=0.2, n_estimators=400;, score=0.970 total time=
1.0s
[CV 3/5] END learning_rate=0.2, n_estimators=400;, score=0.970 total time=
1.1s
[CV 4/5] END learning_rate=0.2, n_estimators=400;, score=0.963 total time=
1.0s
[CV 5/5] END learning_rate=0.2, n_estimators=400;, score=0.969 total time=
1.0s
[CV 1/5] END .learning_rate=1, n_estimators=400;, score=0.939 total time= 0.9s
[CV 2/5] END .learning_rate=1, n_estimators=400;, score=0.942 total time= 0.9s
[CV 3/5] END .learning_rate=1, n_estimators=400;, score=0.938 total time= 1.0s
[CV 4/5] END .learning_rate=1, n_estimators=400;, score=0.929 total time= 0.8s
[CV 5/5] END .learning_rate=1, n_estimators=400;, score=0.936 total time= 1.1s
Fitting 5 folds for each of 10 candidates, totalling 50 fits
[CV 1/5] END .learning_rate=1, n_estimators=800;, score=0.931 total time= 1.5s
[CV 2/5] END .learning_rate=1, n_estimators=800;, score=0.932 total time= 2.0s
[CV 3/5] END .learning_rate=1, n_estimators=800;, score=0.945 total time= 2.2s
[CV 4/5] END .learning_rate=1, n_estimators=800;, score=0.928 total time= 1.4s
[CV 5/5] END .learning_rate=1, n_estimators=800;, score=0.945 total time= 1.4s
[CV 1/5] END learning_rate=0.7, n_estimators=700;, score=0.954 total time=
1.4s
[CV 2/5] END learning_rate=0.7, n_estimators=700;, score=0.941 total time=
1.4s
[CV 3/5] END learning_rate=0.7, n_estimators=700;, score=0.958 total time=
1.4s
[CV 4/5] END learning_rate=0.7, n_estimators=700;, score=0.954 total time=
1.7s
[CV 5/5] END learning_rate=0.7, n_estimators=700;, score=0.954 total time=
1.3s
[CV 1/5] END learning_rate=0.4, n_estimators=100;, score=0.954 total time=
0.6s
[CV 2/5] END learning_rate=0.4, n_estimators=100;, score=0.962 total time=
0.7s
[CV 3/5] END learning_rate=0.4, n_estimators=100;, score=0.961 total time=
0.8s

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```
[CV 4/5] END learning_rate=0.4, n_estimators=100;, score=0.953 total time=
1.0s
[CV 5/5] END learning_rate=0.4, n_estimators=100;, score=0.958 total time=
0.8s
[CV 1/5] END learning_rate=0.2, n_estimators=1000;, score=0.967 total time=
1.8s
[CV 2/5] END learning_rate=0.2, n_estimators=1000;, score=0.970 total time=
2.0s
[CV 3/5] END learning_rate=0.2, n_estimators=1000;, score=0.971 total time=
1.7s
[CV 4/5] END learning_rate=0.2, n_estimators=1000;, score=0.965 total time=
1.7s
[CV 5/5] END learning_rate=0.2, n_estimators=1000;, score=0.969 total time=
2.3s
[CV 1/5] END learning_rate=0.5, n_estimators=800;, score=0.953 total time=
1.5s
[CV 2/5] END learning_rate=0.5, n_estimators=800;, score=0.955 total time=
1.5s
[CV 3/5] END learning_rate=0.5, n_estimators=800;, score=0.960 total time=
1.7s
[CV 4/5] END learning_rate=0.5, n_estimators=800;, score=0.961 total time=
1.5s
[CV 5/5] END learning_rate=0.5, n_estimators=800;, score=0.958 total time=
2.0s
[CV 1/5] END learning_rate=0.4, n_estimators=800;, score=0.958 total time=
1.5s
[CV 2/5] END learning_rate=0.4, n_estimators=800;, score=0.967 total time=
1.7s
[CV 3/5] END learning_rate=0.4, n_estimators=800;, score=0.965 total time=
1.6s
[CV 4/5] END learning_rate=0.4, n_estimators=800;, score=0.959 total time=
1.6s
[CV 5/5] END learning_rate=0.4, n_estimators=800;, score=0.963 total time=
2.1s
[CV 1/5] END learning_rate=0.5, n_estimators=700;, score=0.953 total time=
1.4s
[CV 2/5] END learning_rate=0.5, n_estimators=700;, score=0.955 total time=
1.3s
[CV 3/5] END learning_rate=0.5, n_estimators=700;, score=0.960 total time=
1.3s
[CV 4/5] END learning_rate=0.5, n_estimators=700;, score=0.961 total time=
1.8s
[CV 5/5] END learning_rate=0.5, n_estimators=700;, score=0.958 total time=
1.3s
[CV 1/5] END learning_rate=0.7, n_estimators=100;, score=0.955 total time=
0.5s
[CV 2/5] END learning_rate=0.7, n_estimators=100;, score=0.940 total time=
0.5s
```

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[CV 3/5] END learning_rate=0.7, n_estimators=100;, score=0.957 total time=
0.5s
[CV 4/5] END learning_rate=0.7, n_estimators=100;, score=0.952 total time=
0.4s
[CV 5/5] END learning_rate=0.7, n_estimators=100;, score=0.952 total time=
0.5s
[CV 1/5] END learning_rate=0.2, n_estimators=400;, score=0.966 total time=
0.8s
[CV 2/5] END learning_rate=0.2, n_estimators=400;, score=0.969 total time=
0.8s
[CV 3/5] END learning_rate=0.2, n_estimators=400;, score=0.970 total time=
0.8s
[CV 4/5] END learning_rate=0.2, n_estimators=400;, score=0.963 total time=
0.9s
[CV 5/5] END learning_rate=0.2, n_estimators=400;, score=0.968 total time=
1.5s
[CV 1/5] END learning_rate=0.2, n_estimators=100;, score=0.951 total time=
0.6s
[CV 2/5] END learning_rate=0.2, n_estimators=100;, score=0.956 total time=
0.6s
[CV 3/5] END learning_rate=0.2, n_estimators=100;, score=0.960 total time=
0.5s
[CV 4/5] END learning_rate=0.2, n_estimators=100;, score=0.945 total time=
0.5s
[CV 5/5] END learning_rate=0.2, n_estimators=100;, score=0.951 total time=
0.4s
Fitting 5 folds for each of 10 candidates, totalling 50 fits
[CV 1/5] END .learning_rate=1, n_estimators=800;, score=0.941 total time= 1.3s
[CV 2/5] END .learning_rate=1, n_estimators=800;, score=0.950 total time= 1.7s
[CV 3/5] END .learning_rate=1, n_estimators=800;, score=0.933 total time= 1.3s
[CV 4/5] END .learning_rate=1, n_estimators=800;, score=0.922 total time= 1.3s
[CV 5/5] END .learning_rate=1, n_estimators=800;, score=0.947 total time= 1.4s
[CV 1/5] END learning_rate=0.5, n_estimators=700;, score=0.958 total time=
2.0s
[CV 2/5] END learning_rate=0.5, n_estimators=700;, score=0.966 total time=
1.6s
[CV 3/5] END learning_rate=0.5, n_estimators=700;, score=0.964 total time=
1.4s
[CV 4/5] END learning_rate=0.5, n_estimators=700;, score=0.962 total time=
1.5s
[CV 5/5] END learning_rate=0.5, n_estimators=700;, score=0.965 total time=
1.4s
[CV 1/5] END learning_rate=0.4, n_estimators=800;, score=0.963 total time=
1.6s
[CV 2/5] END learning_rate=0.4, n_estimators=800;, score=0.961 total time=
1.5s
[CV 3/5] END learning_rate=0.4, n_estimators=800;, score=0.972 total time=
1.5s

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[CV 4/5] END learning_rate=0.4, n_estimators=800;, score=0.971 total time=
2.0s
[CV 5/5] END learning_rate=0.4, n_estimators=800;, score=0.962 total time=
1.4s
[CV 1/5] END .learning_rate=1, n_estimators=400;, score=0.941 total time= 1.2s
[CV 2/5] END .learning_rate=1, n_estimators=400;, score=0.950 total time= 1.0s
[CV 3/5] END .learning_rate=1, n_estimators=400;, score=0.933 total time= 1.0s
[CV 4/5] END .learning_rate=1, n_estimators=400;, score=0.922 total time= 1.5s
[CV 5/5] END .learning_rate=1, n_estimators=400;, score=0.947 total time= 1.7s
[CV 1/5] END learning_rate=0.4, n_estimators=400;, score=0.963 total time=
1.1s
[CV 2/5] END learning_rate=0.4, n_estimators=400;, score=0.960 total time=
1.1s
[CV 3/5] END learning_rate=0.4, n_estimators=400;, score=0.972 total time=
1.1s
[CV 4/5] END learning_rate=0.4, n_estimators=400;, score=0.970 total time=
1.3s
[CV 5/5] END learning_rate=0.4, n_estimators=400;, score=0.962 total time=
1.0s
[CV 1/5] END learning_rate=0.4, n_estimators=700;, score=0.963 total time=
1.3s
[CV 2/5] END learning_rate=0.4, n_estimators=700;, score=0.961 total time=
2.1s
[CV 3/5] END learning_rate=0.4, n_estimators=700;, score=0.972 total time=
1.4s
[CV 4/5] END learning_rate=0.4, n_estimators=700;, score=0.971 total time=
1.2s
[CV 5/5] END learning_rate=0.4, n_estimators=700;, score=0.962 total time=
1.4s
[CV 1/5] END learning_rate=0.4, n_estimators=1000;, score=0.963 total time=
1.9s
[CV 2/5] END learning_rate=0.4, n_estimators=1000;, score=0.961 total time=
2.0s
[CV 3/5] END learning_rate=0.4, n_estimators=1000;, score=0.972 total time=
3.5s
[CV 4/5] END learning_rate=0.4, n_estimators=1000;, score=0.971 total time=
2.4s
[CV 5/5] END learning_rate=0.4, n_estimators=1000;, score=0.962 total time=
2.2s
[CV 1/5] END learning_rate=0.7, n_estimators=1000;, score=0.946 total time=
2.0s
[CV 2/5] END learning_rate=0.7, n_estimators=1000;, score=0.962 total time=
4.3s
[CV 3/5] END learning_rate=0.7, n_estimators=1000;, score=0.966 total time=
1.8s
[CV 4/5] END learning_rate=0.7, n_estimators=1000;, score=0.948 total time=
3.0s
[CV 5/5] END learning_rate=0.7, n_estimators=1000;, score=0.947 total time=

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```

1.7s
[CV 1/5] END learning_rate=0.5, n_estimators=100;, score=0.956 total time=
0.6s
[CV 2/5] END learning_rate=0.5, n_estimators=100;, score=0.962 total time=
0.7s
[CV 3/5] END learning_rate=0.5, n_estimators=100;, score=0.961 total time=
0.8s
[CV 4/5] END learning_rate=0.5, n_estimators=100;, score=0.958 total time=
1.1s
[CV 5/5] END learning_rate=0.5, n_estimators=100;, score=0.962 total time=
0.8s
[CV 1/5] END .learning_rate=1, n_estimators=100;, score=0.940 total time= 0.8s
[CV 2/5] END .learning_rate=1, n_estimators=100;, score=0.950 total time= 0.7s
[CV 3/5] END .learning_rate=1, n_estimators=100;, score=0.934 total time= 0.6s
[CV 4/5] END .learning_rate=1, n_estimators=100;, score=0.919 total time= 0.8s
[CV 5/5] END .learning_rate=1, n_estimators=100;, score=0.946 total time= 0.6s
Fitting 5 folds for each of 10 candidates, totalling 50 fits
[CV 1/5] END learning_rate=0.7, n_estimators=700;, score=0.962 total time=
1.5s
[CV 2/5] END learning_rate=0.7, n_estimators=700;, score=0.969 total time=
1.4s
[CV 3/5] END learning_rate=0.7, n_estimators=700;, score=0.957 total time=
1.4s
[CV 4/5] END learning_rate=0.7, n_estimators=700;, score=0.951 total time=
1.5s
[CV 5/5] END learning_rate=0.7, n_estimators=700;, score=0.959 total time=
1.5s
[CV 1/5] END learning_rate=0.7, n_estimators=100;, score=0.960 total time=
0.9s
[CV 2/5] END learning_rate=0.7, n_estimators=100;, score=0.968 total time=
0.7s
[CV 3/5] END learning_rate=0.7, n_estimators=100;, score=0.955 total time=
0.7s
[CV 4/5] END learning_rate=0.7, n_estimators=100;, score=0.950 total time=
0.8s
[CV 5/5] END learning_rate=0.7, n_estimators=100;, score=0.957 total time=
0.6s
[CV 1/5] END .learning_rate=1, n_estimators=800;, score=0.933 total time= 1.6s
[CV 2/5] END .learning_rate=1, n_estimators=800;, score=0.949 total time= 1.9s
[CV 3/5] END .learning_rate=1, n_estimators=800;, score=0.937 total time= 1.5s
[CV 4/5] END .learning_rate=1, n_estimators=800;, score=0.930 total time= 1.5s
[CV 5/5] END .learning_rate=1, n_estimators=800;, score=0.946 total time= 1.6s
[CV 1/5] END learning_rate=0.2, n_estimators=100;, score=0.948 total time=
0.7s
[CV 2/5] END learning_rate=0.2, n_estimators=100;, score=0.959 total time=
0.7s
[CV 3/5] END learning_rate=0.2, n_estimators=100;, score=0.956 total time=
0.7s

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[CV 4/5] END learning_rate=0.2, n_estimators=100;, score=0.949 total time=
0.7s
[CV 5/5] END learning_rate=0.2, n_estimators=100;, score=0.954 total time=
0.7s
[CV 1/5] END learning_rate=1, n_estimators=1000;, score=0.933 total time= 1.7s
[CV 2/5] END learning_rate=1, n_estimators=1000;, score=0.949 total time= 1.8s
[CV 3/5] END learning_rate=1, n_estimators=1000;, score=0.937 total time= 2.2s
[CV 4/5] END learning_rate=1, n_estimators=1000;, score=0.930 total time= 1.9s
[CV 5/5] END learning_rate=1, n_estimators=1000;, score=0.946 total time= 2.2s
[CV 1/5] END .learning_rate=1, n_estimators=100;, score=0.932 total time= 0.9s
[CV 2/5] END .learning_rate=1, n_estimators=100;, score=0.947 total time= 1.0s
[CV 3/5] END .learning_rate=1, n_estimators=100;, score=0.935 total time= 0.8s
[CV 4/5] END .learning_rate=1, n_estimators=100;, score=0.930 total time= 0.6s
[CV 5/5] END .learning_rate=1, n_estimators=100;, score=0.945 total time= 0.8s
[CV 1/5] END learning_rate=0.2, n_estimators=800;, score=0.968 total time=
1.6s
[CV 2/5] END learning_rate=0.2, n_estimators=800;, score=0.974 total time=
1.6s
[CV 3/5] END learning_rate=0.2, n_estimators=800;, score=0.976 total time=
1.7s
[CV 4/5] END learning_rate=0.2, n_estimators=800;, score=0.966 total time=
1.5s
[CV 5/5] END learning_rate=0.2, n_estimators=800;, score=0.973 total time=
2.2s
[CV 1/5] END learning_rate=0.2, n_estimators=400;, score=0.966 total time=
1.2s
[CV 2/5] END learning_rate=0.2, n_estimators=400;, score=0.972 total time=
1.2s
[CV 3/5] END learning_rate=0.2, n_estimators=400;, score=0.974 total time=
1.3s
[CV 4/5] END learning_rate=0.2, n_estimators=400;, score=0.964 total time=
1.1s
[CV 5/5] END learning_rate=0.2, n_estimators=400;, score=0.972 total time=
1.0s
[CV 1/5] END learning_rate=0.4, n_estimators=700;, score=0.958 total time=
1.4s
[CV 2/5] END learning_rate=0.4, n_estimators=700;, score=0.970 total time=
1.7s
[CV 3/5] END learning_rate=0.4, n_estimators=700;, score=0.970 total time=
1.4s
[CV 4/5] END learning_rate=0.4, n_estimators=700;, score=0.955 total time=
1.4s
[CV 5/5] END learning_rate=0.4, n_estimators=700;, score=0.972 total time=
1.5s
[CV 1/5] END learning_rate=0.4, n_estimators=800;, score=0.958 total time=
1.6s
[CV 2/5] END learning_rate=0.4, n_estimators=800;, score=0.970 total time=
1.6s

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[CV 3/5] END learning_rate=0.4, n_estimators=800;, score=0.970 total time=
1.8s
[CV 4/5] END learning_rate=0.4, n_estimators=800;, score=0.955 total time=
1.4s
[CV 5/5] END learning_rate=0.4, n_estimators=800;, score=0.972 total time=
1.6s
Fitting 5 folds for each of 10 candidates, totalling 50 fits
[CV 1/5] END learning_rate=0.2, n_estimators=1000;, score=0.968 total time=
1.9s
[CV 2/5] END learning_rate=0.2, n_estimators=1000;, score=0.970 total time=
2.1s
[CV 3/5] END learning_rate=0.2, n_estimators=1000;, score=0.972 total time=
2.0s
[CV 4/5] END learning_rate=0.2, n_estimators=1000;, score=0.968 total time=
1.7s
[CV 5/5] END learning_rate=0.2, n_estimators=1000;, score=0.975 total time=
2.2s
[CV 1/5] END .learning_rate=1, n_estimators=700;, score=0.951 total time= 1.5s
[CV 2/5] END .learning_rate=1, n_estimators=700;, score=0.934 total time= 1.5s
[CV 3/5] END .learning_rate=1, n_estimators=700;, score=0.937 total time= 1.4s
[CV 4/5] END .learning_rate=1, n_estimators=700;, score=0.949 total time= 1.4s
[CV 5/5] END .learning_rate=1, n_estimators=700;, score=0.937 total time= 1.4s
[CV 1/5] END learning_rate=0.7, n_estimators=800;, score=0.936 total time=
2.7s
[CV 2/5] END learning_rate=0.7, n_estimators=800;, score=0.945 total time=
1.8s
[CV 3/5] END learning_rate=0.7, n_estimators=800;, score=0.966 total time=
1.5s
[CV 4/5] END learning_rate=0.7, n_estimators=800;, score=0.958 total time=
1.6s
[CV 5/5] END learning_rate=0.7, n_estimators=800;, score=0.961 total time=
3.1s
[CV 1/5] END learning_rate=0.7, n_estimators=100;, score=0.934 total time=
0.7s
[CV 2/5] END learning_rate=0.7, n_estimators=100;, score=0.945 total time=
0.7s
[CV 3/5] END learning_rate=0.7, n_estimators=100;, score=0.965 total time=
0.8s
[CV 4/5] END learning_rate=0.7, n_estimators=100;, score=0.956 total time=
0.8s
[CV 5/5] END learning_rate=0.7, n_estimators=100;, score=0.959 total time=
0.7s
[CV 1/5] END learning_rate=0.7, n_estimators=700;, score=0.937 total time=
1.4s
[CV 2/5] END learning_rate=0.7, n_estimators=700;, score=0.945 total time=
1.4s
[CV 3/5] END learning_rate=0.7, n_estimators=700;, score=0.966 total time=
1.6s

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[CV 4/5] END learning_rate=0.7, n_estimators=700;, score=0.958 total time=
1.5s
[CV 5/5] END learning_rate=0.7, n_estimators=700;, score=0.961 total time=
1.9s
[CV 1/5] END .learning_rate=1, n_estimators=100;, score=0.952 total time= 0.9s
[CV 2/5] END .learning_rate=1, n_estimators=100;, score=0.934 total time= 0.9s
[CV 3/5] END .learning_rate=1, n_estimators=100;, score=0.936 total time= 0.7s
[CV 4/5] END .learning_rate=1, n_estimators=100;, score=0.945 total time= 0.7s
[CV 5/5] END .learning_rate=1, n_estimators=100;, score=0.938 total time= 0.8s
[CV 1/5] END learning_rate=0.2, n_estimators=800;, score=0.968 total time=
1.7s
[CV 2/5] END learning_rate=0.2, n_estimators=800;, score=0.970 total time=
1.5s
[CV 3/5] END learning_rate=0.2, n_estimators=800;, score=0.972 total time=
1.5s
[CV 4/5] END learning_rate=0.2, n_estimators=800;, score=0.967 total time=
1.6s
[CV 5/5] END learning_rate=0.2, n_estimators=800;, score=0.975 total time=
1.7s
[CV 1/5] END learning_rate=0.4, n_estimators=700;, score=0.951 total time=
1.4s
[CV 2/5] END learning_rate=0.4, n_estimators=700;, score=0.967 total time=
1.5s
[CV 3/5] END learning_rate=0.4, n_estimators=700;, score=0.966 total time=
1.9s
[CV 4/5] END learning_rate=0.4, n_estimators=700;, score=0.966 total time=
1.5s
[CV 5/5] END learning_rate=0.4, n_estimators=700;, score=0.967 total time=
1.4s
[CV 1/5] END learning_rate=0.2, n_estimators=100;, score=0.949 total time=
0.8s
[CV 2/5] END learning_rate=0.2, n_estimators=100;, score=0.953 total time=
0.9s
[CV 3/5] END learning_rate=0.2, n_estimators=100;, score=0.953 total time=
0.7s
[CV 4/5] END learning_rate=0.2, n_estimators=100;, score=0.952 total time=
0.7s
[CV 5/5] END learning_rate=0.2, n_estimators=100;, score=0.956 total time=
0.7s
[CV 1/5] END learning_rate=0.4, n_estimators=100;, score=0.950 total time=
0.5s
[CV 2/5] END learning_rate=0.4, n_estimators=100;, score=0.963 total time=
0.7s
[CV 3/5] END learning_rate=0.4, n_estimators=100;, score=0.963 total time=
0.9s
[CV 4/5] END learning_rate=0.4, n_estimators=100;, score=0.957 total time=
0.7s
[CV 5/5] END learning_rate=0.4, n_estimators=100;, score=0.958 total time=

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0.6s
Average cross validation R2 score: 0.9753825201956607
[Parallel(n_jobs=1)]: Done    8 out of  8 | elapsed: 10.0min finished
[142]: joblib.dump(grid_cat, 'walmart_sales_predictor.h5')
[142]: ['walmart_sales_predictor.h5']
[143]: model = joblib.load('walmart_sales_predictor.h5')
model
[143]: RandomizedSearchCV(cv=5,
                          estimator=<catboost.core.CatBoostRegressor object at
0x000001BE0B867DF0>,
                          param_distributions={'learning_rate': [0.2, 0.4, 0.5, 0.7,
                                                               1],
                                               'n_estimators': [100, 400, 700, 800,
                                                               1000]},  
verbose=5)
```