



# ML Visualization Cheat Sheet



## Tools

With the help of following Python libraries, it makes it possible to understand ML data with statistics

### YELLOWBRICK

- For Learning Curve:

```
from yellowbrick.model_selection import learning_curve
```

- For Validation Curve:

```
from yellowbrick.model_selection import validation_curve
```

- For Precision-Recall Curve:

```
from yellowbrick.classifier import PrecisionRecallCurve
```

### SKLEARN

- For ROC Curve:

```
from sklearn.metrics import roc_curve
```

- For Confusion Matrix:

```
from sklearn.metrics import confusion_matrix
```

- For Precision-Recall Curve:

```
from sklearn.linear_model import LinearRegression
```

### SEABORN

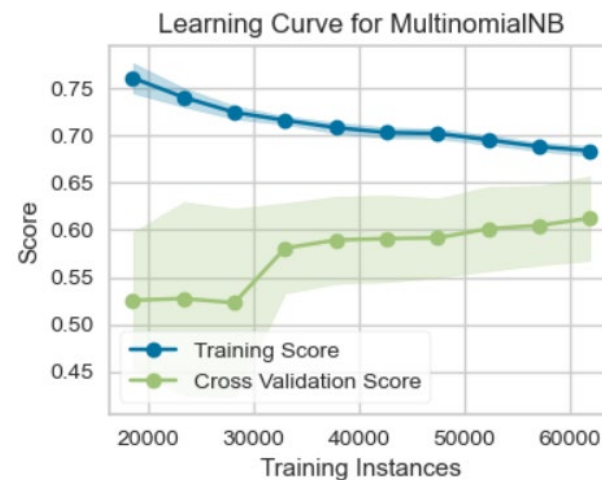
- For Confusion and Correlation Matrices:

```
import seaborn as sns
```

## Classification

### LEARNING CURVES

DO WE HAVE ENOUGH DATA?



```
learning_curve(MultinomialNB(), X, y)
```

Where

- MultinomialNB() – classifier of your choice
- X, y – classification dataset

### VALIDATON CURVES

WHAT ARE THE OPTIMAL HYPERPARAMETERS?



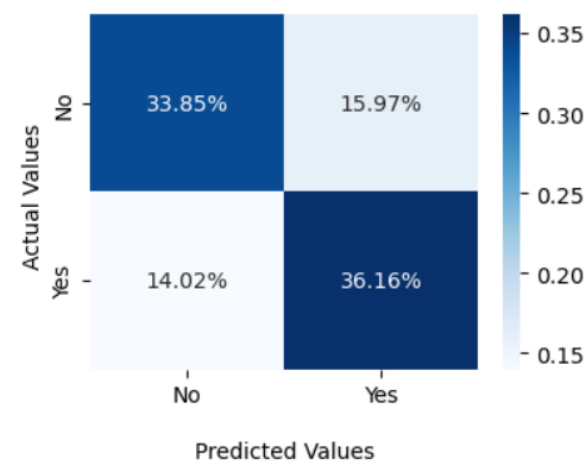
```
validation_curve(DecisionTreeRegressor(),  
param_name="max_depth",  
param_range=np.arange(1, 11))
```

Where

- DecisionTreeRegressor() – classifier of your choice
- param\_name – name of the varied parameter

### CONFUSION MATRIX

WHICH CLASSES ARE MIXED-UP?



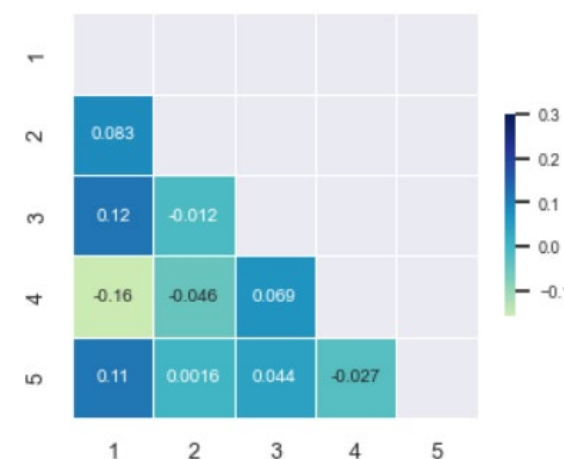
```
cm = confusion_matrix(y_test, y_pred)  
sns.heatmap(cm/np.sum(cm), fmt='.2%', annot=True,  
cmap='Blues')
```

Where

- y\_test, y\_pred – test set, training set

### CORRELATION MATRIX

HOW CORRELATED ARE PARAMETERS?



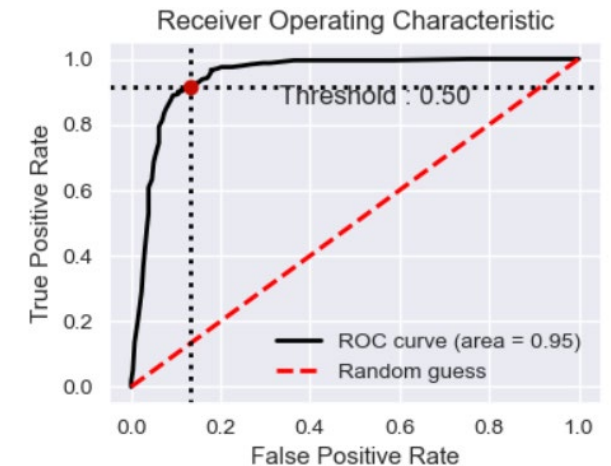
```
mask = np.triu(np.ones_like(dataset, dtype=bool))  
sns.heatmap(dataset, mask=mask, cmap="YlGnBu",  
annot=True, vmax=.3, center=0,  
square=True, linewidths=.5, cbar_kws={"shrink": .5})
```

Where

- dataset – used dataset

### ROC CURVE

IS CLASSIFIER A GOOD RANKER?



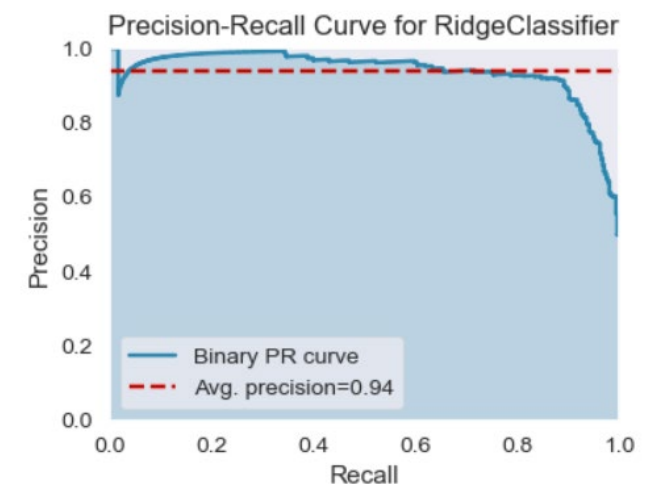
```
bc=BinaryClassification(y_test, y_pred, labels=[c1, c2])  
bc.plot_roc_curve()
```

Where

- y\_test, y\_pred – test set, training set
- BinaryClassification() – classification of your choice

### PRECISION-RECALL CURVE

WHAT IS THE PRECISION-RECALL TRADEOFF?



```
p = PrecisionRecallCurve(RidgeClassifier(random_state=0))  
p.fit(X_train, y_train)  
p.score(X_test, y_test)
```

Where

- X\_train, X\_test, y\_test, y\_pred – test sets, training sets

## SHAP

- For Feature Importance Plot and SHAP summaries:

```
import xgboost
import shap
X, y = shap.dataset
model = xgboost.XGBRegressor().fit(X, y)
explainer = shap.Explainer(model)
```

Where

- model – an XGBoost model to train
- X, y – a dataset to train
- dataset – used dataset

## YELLOWBRICK

- For Intercluster Distance Map:

```
from yellowbrick.cluster import
intercluster_distance
```

- For Residuals Plot:

```
from yellowbrick.regressor import
residuals_plot
```

- For Principal Component Plot:

```
from yellowbrick.features import PCA
```

## PMDARIMA

- For Seasonality Decomposition:

```
from pmdarima import arima
from pmdarima import utils
```

## USEFUL LINKS

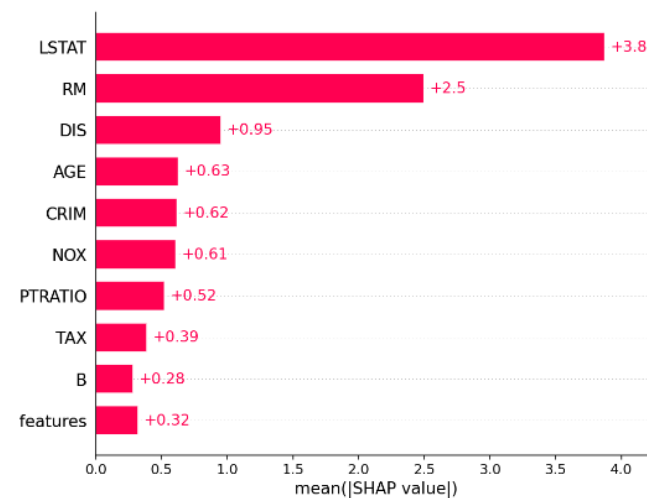
- [SHAP](#)
- [YELLOWBRICK](#)
- [PMDARIMA](#)
- [SKLEARN](#)
- [SEABORN](#)



## Feature Analysis

### FEATURE IMPORTANCE PLOT

MOST IMPORTANT PREDICTION  
FEATURES



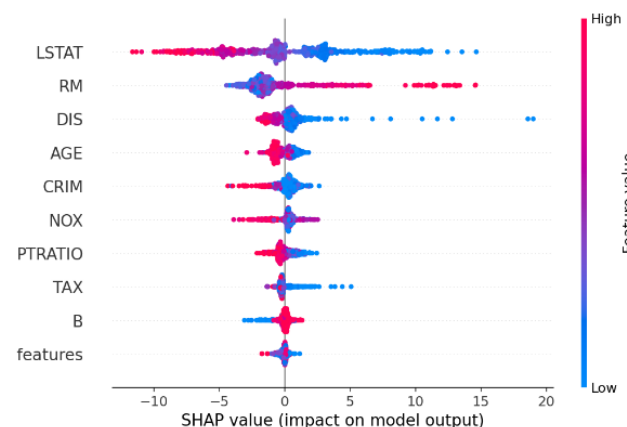
```
shap_values = explainer(X)
shap.plots.bar(shap_values)
```

Where

- shap\_values – model's prediction explanation

### SHAP SUMMARIES

FEATURE IMPORTANCES FOR EACH  
PREDICTION



```
shap_values = explainer(X)
shap.plots.bar(shap_values)
```

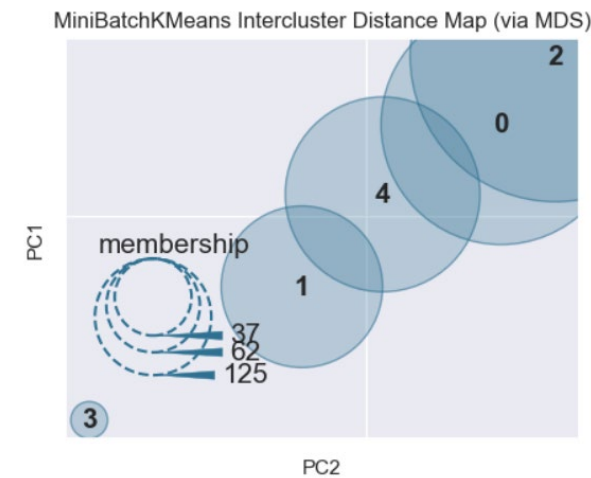
Where

- shap\_values – model's prediction explanation

## Clustering

### INTERCLUSTER DISTANCE MAP

HOW DISTANT ARE THE CLUSTERS?



```
intercluster_distance(MiniBatchKMeans(5,
random_state=777), X)
```

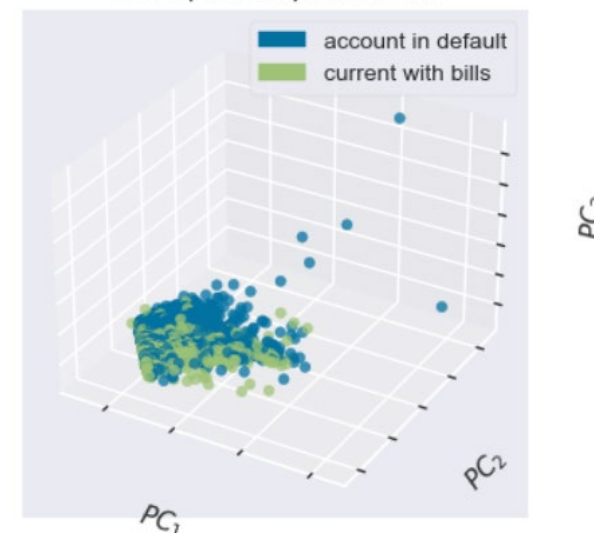
Where

- MiniBatchKMeans – used algorithm

### PRINCIPAL COMPONENT PLOT

3D-LOOK OF DATASET

Principal Component Plot



```
p = PCA(scale=True, projection=3, classes=classes)
p.fit_transform(X, y)
```

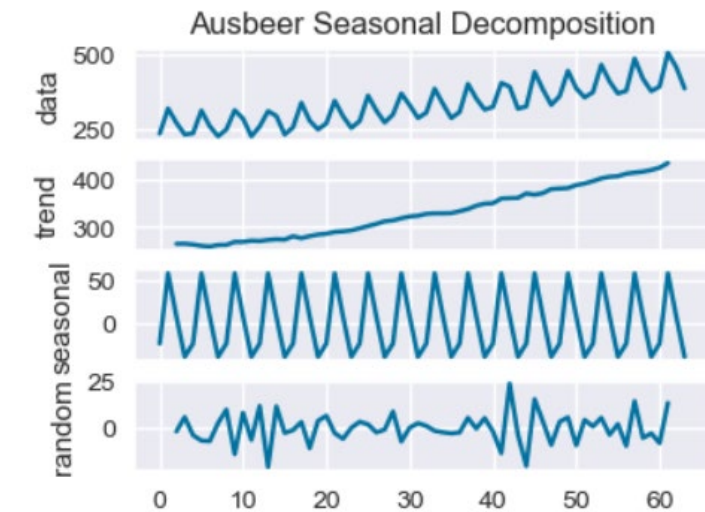
Where

- X, y – used dataset
- classes – class labels

## Regression & Time Series

### SEASONALITY DECOMPOSITION

SEASON, TRADE, NOISE IM TIME  
SERIES



```
utils.decomposed_plot(decomposed,
figure_kwargs=figure_kwargs, show=False)
```

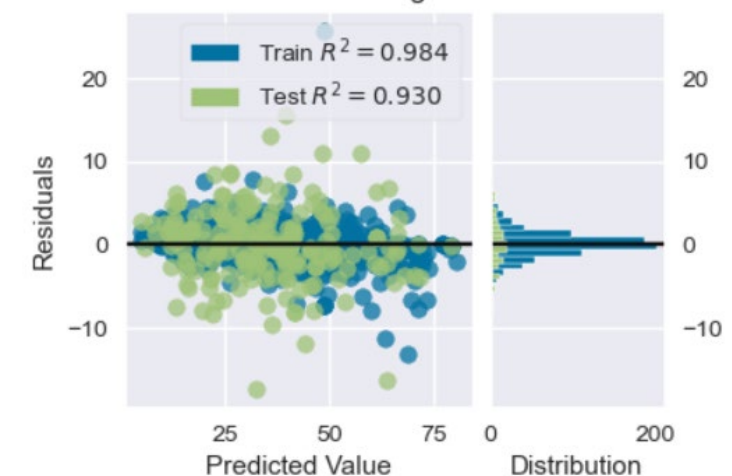
Where

- decomposed – tuple of y-axis variable datasets

### RESIDUALS PLOT

ARE REGRESSION ERRORS  
NORMALLY DISTRIBUTED?

Residuals for RandomForestRegressor Model



```
residuals_plot(RandomForestRegressor(), X_train,
y_train, X_test, y_test)
```

Where

- X\_train, X\_test, y\_train, y\_test – test sets, training sets
- RandomForestRegressor() – estimator of your choice