

### Introduction

This cheat sheet provides a comparison between basic data processing technique as well as machine learning models in both R and Python.

### Documentations

[https://scikit-learn.org/stable/auto\\_examples/index.html](https://scikit-learn.org/stable/auto_examples/index.html)  
<https://seaborn.pydata.org/>  
<https://cran.r-project.org/web/packages/rpart/index.html>  
<https://cran.r-project.org/web/packages/caret/index.html>  
<https://cran.r-project.org/web/packages/randomForest/index.html>  
<https://www.rdocumentation.org/packages/stats/versions/3.6.2>

### Load dataset in R

<code>library(datasets)</code>	Import packages
<code>data(iris)</code>	Load dataset
<code>head(iris)</code>	Look up the first 6 rows of the dataset
<code>summary(iris)</code>	Get summary statistics of each columns
<code>names(iris)</code>	Get the column names

### Data preprocessing in R

<code>scaling = preProcess(data, method = c('center', 'scale'))</code>	Create scaling based on data
<code>data_scaled = predict(scaling, data)</code>	Apply scaling to data
<code>train_partition = createDataPartition(y, p = 0.8, list = FALSE)</code>	Balanced splitting based on the outcome ( 80/20 split)
<code>data_train = data[train_partition,]</code>	Split data into train and test sets
<code>data_test = data[-train_partition,]</code>	Split data into train and test sets

### Supervised learning models in R

<code>model = lm(data, y ~ x)</code>	Simple linear regression
<code>model = lm(data, y ~ x1 + x2 + x3)</code>	Multiple linear regression
<code>summary(model)</code>	Print summary statistics from linear model
<code>predictions = predict(object, newdata)</code>	Make prediction based on the model object
<code>model = glm(data, y ~ x1 + x2 + x3, family = 'binomial')</code>	Logistic regression

### Supervised learning models in R (cont)

<code>model = svm(data, y ~ x1 + x2 + x3, params)</code>	Support vector machines (SVM)
<code>model = rpart(data, y ~ x1 + x2 + x3, params)</code>	Decision trees
<code>model = randomForest(data, y ~ x1 + x2 + x3, params)</code>	Random forest
<code>data_xgb = xgb.DMatrix(data, label)</code>	Transform the data into DMatrix format
<code>model = xgb.train(data_xgb, label, params)</code>	Gradient boosting models
<code>predictions = knn(train, test, cl, params)</code>	k-NN with labels cl and parameters (e.g., number of neighbors)

### Unsupervised learning models

<code>model = kmeans(x, params)</code>	K-Means clustering
<code>model = prcomp(x, params)</code>	Principal components analysis (PCA)

### Model performance in R

<code>RMSE(pred, actual)</code>	Root mean square error
<code>R2(pred, actual, form = 'traditional')</code>	Proportion of the variance explained by the model
<code>mean(actual == pred)</code>	Accuracy (how accurate positive predictions are)
<code>confusionMatrix(actual, pred)</code>	Confusion matrix
<code>auc(actual, pred)</code>	Area under the ROC curve
<code>f1Score(actual, pred)</code>	Harmonic mean of precision and recall

### Data visualization in R

<code>geom_point(x, y, color, size, fill, alpha)</code>	Scatter plot
<code>geom_line(x, y, color, size, fill, alpha, linetype)</code>	Line plot
<code>geom_bar(x, y, color, size, fill, alpha)</code>	Bar chart
<code>geom_boxplot(x, y, color)</code>	Box plot
<code>geom_tile(x, y, color, fill)</code>	Heatmap

### Import file in Python

<code>import pandas as pd</code>	Import package
<code>df = pd.read_csv()</code>	Read csv files
<code>df.head(n)</code>	Look up the first n rows of the dataset
<code>df.describe()</code>	Get summary statistics of each columns
<code>df.columns</code>	Get column names

### Data Processing in Python

<code>X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=0)</code>	Split the dataset into training (80%) and test (20%) sets
<code>scaler = StandardScaler()</code>	Standardize features by removing the mean and scaling to unit variance
<code>X_train = scaler.fit_transform(X_train)</code>	Fit and transform scalar on X_train
<code>X_test = scaler.transform(X_test)</code>	Transform X_test

### Supervised learning models in Python

<code>model = LinearRegression()</code>	Linear regression
<code>model.fit(X_train, y_train)</code>	Fit linear model
<code>model.predict(X_test)</code>	Predict using the linear model
<code>LogisticRegression().fit(X_train, y_train)</code>	Logistic regression
<code>LinearSVC.fit(X_train, y_train)</code>	Train primal SVM
<code>SVC().fit(X_train, y_train)</code>	Train dual SVM
<code>DecisionTreeClassifier().fit(X_train, y_train)</code>	Decision tree classifier
<code>RandomForestClassifier().fit(X_train, y_train)</code>	Random forest classifier
<code>GradientBoostingClassifier().fit(X_train, y_train)</code>	Gradient boosting for classification
<code>XGBClassifier().fit(X_train, y_train)</code>	XGboost classifier
<code>KNeighborsClassifier().fit(X_train, y_train)</code>	k-NN

### Unsupervised learning models

<code>KMeans().fit(X)</code>	K-Means clustering
<code>PCA().fit(X)</code>	Principal component analysis (PCA)

### Model performance in Python

<code>metrics.mean_squared_error(y_true, y_pred, squared=False)</code>	Root mean squared error
<code>metrics.r2_score(y_true, y_pred)</code>	Proportion of the variance explained by the model
<code>metrics.confusion_matrix(y_true, y_pred)</code>	Confusion matrix
<code>metrics.accuracy_score(y_true, y_pred)</code>	Accuracy classification score
<code>metrics.roc_auc_score()</code>	Compute ROC-AUC from prediction scores
<code>f1_score(y_true, y_pred, average='macro')</code>	Harmonic mean of the precision and recall

### Data visualization in Python

<code>sns.scatterplot(x, y, hue, size)</code>	Scatter plot
<code>sns.lineplot(x, y, hue, size)</code>	Line plot
<code>sns.barplot(x, y, hue)</code>	Bar chart
<code>sns.boxplot(x, y, hue)</code>	Box plot
<code>sns.heatmap(data, linecolor, linewidth)</code>	Heatmap

