```
In [86]:
         # packages
          %matplotlib notebook
          import numpy as np
          import pandas as pd
          from collections import Counter
          import warnings
          warnings.filterwarnings("ignore")
In [87]: | df = pd.read csv("zoo.csv")
In [88]: df.info()
          <class 'pandas.core.frame.DataFrame'>
          RangeIndex: 101 entries, 0 to 100
          Data columns (total 18 columns):
                             Non-Null Count
               Column
                                               Dtype
           0
                             101 non-null
               animal name
                                               object
           1
               hair
                             101 non-null
                                               int64
           2
               feathers
                             101 non-null
                                               int64
           3
                             101 non-null
                                               int64
               eggs
           4
               milk
                             101 non-null
                                               int64
           5
               airborne
                             101 non-null
                                               int64
           6
               aquatic
                             101 non-null
                                               int64
           7
               predator
                             101 non-null
                                               int64
           8
               toothed
                             101 non-null
                                               int64
           9
               backbone
                             101 non-null
                                               int64
           10
               breathes
                             101 non-null
                                               int64
           11
               venomous
                             101 non-null
                                               int64
           12
               fins
                             101 non-null
                                               int64
           13
               legs
                             101 non-null
                                               int64
           14
              tail
                             101 non-null
                                               int64
           15
               domestic
                             101 non-null
                                               int64
           16
                             101 non-null
                                               int64
              catsize
               class_type
                             101 non-null
                                               int64
          dtypes: int64(17), object(1)
          memory usage: 14.3+ KB
In [89]: df.head()
Out[89]:
              animal_name
                          hair feathers eggs milk
                                                  airborne aquatic predator toothed
                                                                                   backbone
                                                                                             breat
           0
                  aardvark
                            1
                                               1
                                                        0
                                                                0
                                                                         1
                                                                                1
                                                                                          1
           1
                                    0
                                          0
                                               1
                                                        0
                                                                0
                                                                        0
                                                                                1
                                                                                          1
                  antelope
                            1
           2
                                    0
                                               0
                                                        0
                                                                1
                                                                         1
                     bass
                            0
                                                                                          1
           3
                     bear
                            1
                                     0
                                          0
                                               1
                                                        0
                                                                0
                                                                         1
                                                                                1
                                                                                          1
```

localhost:8888/notebooks/Predict Zoo lab 4	.ipynb
--	--------

1

boar

0

0

1

0

0

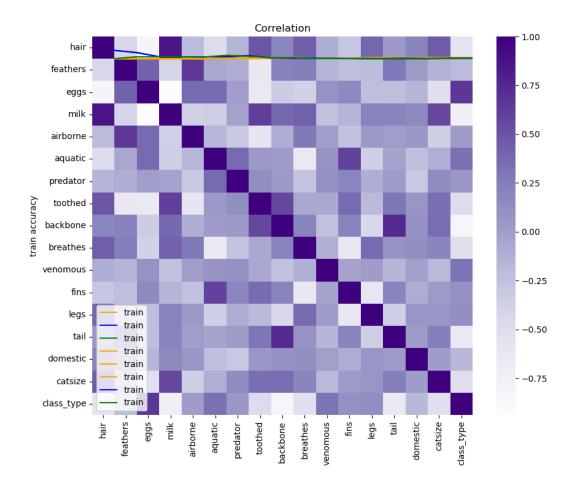
1

1

1

```
In [90]: #correlation value between features
    import seaborn as sns
    import matplotlib.pyplot as plt
    corr = ds.corr()
    fig = plt.figure(figsize=(10,8))
    r = sns.heatmap(corr, cmap='Purples')
    r.set_title("Correlation")
```

<IPython.core.display.Javascript object>



Out[90]: Text(0.5, 1.0, 'Correlation')

```
In [91]: # data preperation
y = ds["class_type"].values
x_ds=ds.drop(["animal_name"],axis=1)
x = (x_ds-np.min(x_ds))/(np.max(x_ds)-np.min(x_ds))
from sklearn.model_selection import train_test_split
x_tn, x_ts, y_tn, y_ts = train_test_split(x,y,test_size = 0.2,random_state=1)
```

In [92]: x\_tn

Out[92]:		hair	feathers	eggs	milk	airborne	aquatic	predator	toothed	backbone	breathes	venomoι
	32	1.0	0.0	0.0	1.0	0.0	0.0	0.0	1.0	1.0	1.0	0
	40	1.0	0.0	1.0	0.0	1.0	0.0	0.0	0.0	0.0	1.0	0
	39	1.0	0.0	1.0	0.0	1.0	0.0	0.0	0.0	0.0	1.0	1
	38	0.0	0.0	1.0	0.0	0.0	1.0	1.0	1.0	1.0	0.0	0
	46	0.0	0.0	1.0	0.0	0.0	1.0	1.0	0.0	0.0	0.0	0
	75	1.0	0.0	0.0	1.0	0.0	1.0	1.0	1.0	1.0	1.0	0
	9	1.0	0.0	0.0	1.0	0.0	0.0	0.0	1.0	1.0	1.0	0
	72	0.0	0.0	0.0	0.0	0.0	0.0	1.0	0.0	0.0	1.0	1
	12	0.0	0.0	1.0	0.0	0.0	1.0	1.0	1.0	1.0	0.0	0

1.0

0.0

1.0

0.0

1.0

1.0

80 rows × 17 columns

1.0

1.0 0.0

**37** 0.0

```
In [93]: # KNN package
         from sklearn.neighbors import KNeighborsClassifier
         # Create KNN Classifier
         knn = KNeighborsClassifier(n_neighbors=2 )
In [94]: # Train the model using the training sets
         knn.fit(x_tn, y_tn)
Out[94]: KNeighborsClassifier(n_neighbors=2)
In [95]: # accuracy or score #train
         knn.score(x_tn, y_tn)
Out[95]: 0.9625
In [96]: # accuracy or score #test
         knn.score(x_ts, y_ts)
Out[96]: 0.9047619047619048
In [97]: from sklearn.linear_model import LogisticRegression
         # creating linear regression object.
         lgrgmodel = LogisticRegression()
```

```
In [98]: # Train the model using the train sets.
          lgrgmodel.fit(x_tn,y_tn)
 Out[98]: LogisticRegression()
 In [99]: # score of LR (Accuracy) #train
          lgrgmodel.score(x_tn, y_tn)
 Out[99]: 0.975
In [100]: # score of LR (Accuracy) #test
          lgrgmodel.score(x_ts, y_ts)
Out[100]: 0.9523809523809523
In [101]: # SVM package
          from sklearn.svm import SVC
          svm=SVC(random state=1)
In [102]: #train model using train set
          svm.fit(x_tn,y_tn)
Out[102]: SVC(random_state=1)
In [103]: # calc score of SVM (Accuracy)
          print("train accuracy:",svm.score(x_tn,y_tn))
          print("test accuracy:",svm.score(x_ts,y_ts))
          train accuracy: 0.9875
          test accuracy: 0.9047619047619048
  In [ ]: from sklearn.feature_selection import SelectKBest
          from sklearn.feature_selection import f_classif
  In [ ]:
```

```
In [115]:
          #improve the accuracy of prediction.#KNN
          print("=======#KNN")
          accuracy list train KNN = []
          number of features KNN = np.arange(1,18,1)
          for each in number of features KNN:
             x_new_KNN = SelectKBest(f_classif, k = each).fit_transform(x_tn, y_tn)
             knn.fit (x new KNN, y tn)
             accuracy_list_train_KNN.append(knn.score(x_new_KNN, y_tn))
          plt.plot(number_of_features_KNN,accuracy_list_train_KNN,color="orange", label="tr
          plt.xlabel("number of features")
          plt.ylabel("train accuracy")
          plt.legend()
          plt.show()
          #improve the accuracy of prediction.#LR
          print("=======#LR")
          accuracy list train LR = []
          number of features LR = np.arange(1,18,1)
          for each in number of features LR:
             x_new_LR = SelectKBest(f_classif, k = each).fit_transform(x_tn, y_tn)
             lgrgmodel.fit(x new LR, y tn)
             accuracy list train LR.append(lgrgmodel.score(x new LR, y tn))
          plt.plot(number_of_features_LR,accuracy_list_train_LR,color="blue", label="train'
          plt.xlabel("number of features")
          plt.ylabel("train accuracy")
          plt.legend()
          plt.show()
          #improve the accuracy of prediction. #SVM
          print("=======#SVM")
          accuracy list train = []
          number of features = np.arange(1,18,1)
          for each in number of features:
             x new = SelectKBest(f classif, k = each).fit transform(x tn, y tn)
             svm. fit (x_new, y_tn)
             accuracy list train.append(svm.score(x new, y tn))
          plt.plot(number of features,accuracy list train,color="green", label="train")
          plt.xlabel("number of features")
          plt.ylabel("train accuracy")
          plt.legend()
          plt.show()
```

```
========#KNN
=========#LR
=========#SVM
```

```
In [105]: #best accuracy
          #KNN
          print("=======#KNN")
          d_KNN = {'best features number': number_of_features_KNN, 'train_score': accuracy
          df KNN = pd.DataFrame(data=d KNN)
          print("max accuracy:",df KNN["train score"].max())
          print("max accuracy id:",df_KNN["train_score"].idxmax() )
          #LR
          print("=======#LR")
          d LR = {'best features number': number of features LR, 'train score': accuracy li
          df LR = pd.DataFrame(data=d LR)
          print("max accuracy:",df LR["train score"].max())
          print("max accuracy id:",df_LR["train_score"].idxmax() )
          #SVM
          print("=======#SVM")
          d = {'best features number': number_of_features, 'train_score': accuracy_list_train_score': accuracy_list_train_score'
          df = pd.DataFrame(data=d)
          print("max accuracy:",df["train score"].max())
          print("max accuracy id:",df["train_score"].idxmax() )
```

```
In [106]:
         #KNN
         print("=======#KNN")
         print("max accuracy values: \n",df KNN.iloc[9])
         #LR
         print("=======#LR")
         print("max accuracy values: \n",df LR.iloc[9])
         #SVM
         print("=======#SVM")
         print("max accuracy values: \n",df.iloc[9])
         ========#KNN
         max accuracy values:
          best features number
                                10.0000
                                0.9875
         train score
         Name: 9, dtype: float64
         ========#LR
         max accuracy values:
          best features number
                                10.000
                                0.975
         train_score
         Name: 9, dtype: float64
         ==========================#SVM
         max accuracy values:
          best features number
                                10.0000
                                0.9875
         train score
         Name: 9, dtype: float64
In [107]: # Arrange the train and test dataset including best features
         #KNN
         selector = SelectKBest(f_classif, k = 10)
         x new KNN = selector.fit transform(x tn, y tn)
         x_new_test_KNN=selector. fit_transform(x_ts, y_ts)
         names train KNN = x tn.columns.values[selector.get support()]
         names test KNN = x ts.columns.values[selector.get support() ]
         print("x train features:",names train KNN)
         print("x test features:",names_test_KNN)
         x train features: ['hair' 'feathers' 'eggs' 'milk' 'airborne' 'toothed' 'backbo
         ne'
          'breathes' 'tail' 'class type']
         x test features: ['hair' 'feathers' 'eggs' 'milk' 'airborne' 'toothed' 'backbon
          'breathes' 'tail' 'class_type']
```

```
In [108]: # Arrange the train and test dataset including best features
          #LR
          selector = SelectKBest(f classif, k = 10)
          x_new_LR = selector.fit_transform(x_tn, y_tn)
          x new test LR=selector. fit transform(x ts, y ts)
          names train LR = x tn.columns.values[selector.get support()]
          names_test_LR = x_ts.columns.values[selector.get_support() ]
          print("x train features:",names_train_LR)
          print("x test features:",names_test_LR)
          x train features: ['hair' 'feathers' 'eggs' 'milk' 'airborne' 'toothed' 'backbo
          ne'
           'breathes' 'tail' 'class type']
          x test features: ['hair' 'feathers' 'eggs' 'milk' 'airborne' 'toothed' 'backbon
          e'
           'breathes' 'tail' 'class type']
In [109]: # Arrange the train and test dataset including best features
          #SVM
          selector = SelectKBest(f classif, k = 10)
          x_new = selector.fit_transform(x_tn, y_tn)
          x_new_test=selector. fit_transform(x_ts, y_ts)
          names train = x tn.columns.values[selector.get support()]
          names test = x ts.columns.values[selector.get support() ]
          print("x train features:",names_train)
          print("x test features:",names_test)
          x train features: ['hair' 'feathers' 'eggs' 'milk' 'airborne' 'toothed' 'backbo
          ne'
            'breathes' 'tail' 'class_type']
          x test features: ['hair' 'feathers' 'eggs' 'milk' 'airborne' 'toothed' 'backbon
          e'
            'breathes' 'tail' 'class_type']
```

```
In [110]: #Re-train and re-calculate the model accuracy using the new arrangement of feature
          #import backage
          from sklearn.neighbors import KNeighborsClassifier
          from sklearn import metrics
          knn = KNeighborsClassifier( )
          knn.fit(x new, y tn)
          #Calc score or (Accuracy)
          print("train accuracy: ",knn.score(x_new_KNN, y_tn))
          print("test accuracy: ",knn.score(x_new_test_KNN,y_ts))
          train accuracy: 0.975
          test accuracy: 0.9047619047619048
In [111]: #Re-train and re-calculate the model accuracy using the new arrangement of featur
          #LR
          #import backage
          from sklearn.linear model import LogisticRegression
          lgrgmodel = LogisticRegression()
          lgrgmodel.fit(x_new, y_tn)
          #Calc score or (Accuracy)
          print("train accuracy: ",lgrgmodel.score(x new LR, y tn))
          print("test accuracy: ",lgrgmodel.score(x_new_test_LR,y_ts))
          train accuracy: 0.975
          test accuracy: 0.8571428571428571
In [112]: #Re-train and re-calculate the model accuracy using the new arrangement of feature
          #SVM
          from sklearn.svm import SVC
          svm=SVC( random state=1)
          svm. fit (x_new, y_tn)
          print("train accuracy:",svm.score(x_new, y_tn) )
          print("test accuracy:",svm.score(x_new_test,y_ts) )
```

train accuracy: 0.9875