

In [1]:

```
import pandas as pd
import numpy as np
from sklearn.model_selection import KFold
from sklearn.model_selection import cross_val_score
from sklearn.neighbors import KNeighborsClassifier
import warnings
warnings.filterwarnings("ignore")
```

In [3]:

```
zoo = pd.read_csv('C:/Users/acer/Sandesh Pal/Data Science Assgn/KNN/Zoo.csv')
```

In [4]:

```
zoo.head()
```

Out[4]:

	animal name	hair	feathers	eggs	milk	airborne	aquatic	predator	toothed	backbone	breathes	venomous	fins	legs	tail	domestic	catsize	t
0	aardvark	1	0	0	1	0	0	1	1	1	1	0	0	4	0	0	1	
1	antelope	1	0	0	1	0	0	0	1	1	1	0	0	4	1	0	1	
2	bass	0	0	1	0	0	1	1	1	1	0	0	1	0	1	0	0	
3	bear	1	0	0	1	0	0	1	1	1	1	0	0	4	0	0	1	
4	boar	1	0	0	1	0	0	1	1	1	1	0	0	4	1	0	1	

```
# checking for null values
zoo.info()
```

In [5]:

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 101 entries, 0 to 100
Data columns (total 18 columns):
#   Column          Non-Null Count  Dtype
---  -
0   animal name     101 non-null   object
1   hair            101 non-null   int64
2   feathers        101 non-null   int64
3   eggs            101 non-null   int64
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  catsize         101 non-null   int64
17  type            101 non-null   int64
dtypes: int64(17), object(1)
memory usage: 14.3+ KB
```

In [6]:

```
zoo.describe()
```

```

Out[6]:

```

	hair	feathers	eggs	milk	airborne	aquatic	predator	toothed	backbone	breathes	venomous
count	101.000000	101.000000	101.000000	101.000000	101.000000	101.000000	101.000000	101.000000	101.000000	101.000000	101.000000
mean	0.425743	0.198020	0.584158	0.405941	0.237624	0.356436	0.554455	0.603960	0.821782	0.792079	0.079207
std	0.496921	0.400495	0.495325	0.493522	0.427750	0.481335	0.499505	0.491512	0.384605	0.407844	0.271410
min	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
25%	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	1.000000	1.000000	0.000000
50%	0.000000	0.000000	1.000000	0.000000	0.000000	0.000000	1.000000	1.000000	1.000000	1.000000	0.000000
75%	1.000000	0.000000	1.000000	1.000000	0.000000	1.000000	1.000000	1.000000	1.000000	1.000000	0.000000
max	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000

```

In [7]:
X = zoo.drop(['type','animal name'], axis =1)
X.head()

```

```

Out[7]:

```

	hair	feathers	eggs	milk	airborne	aquatic	predator	toothed	backbone	breathes	venomous	fins	legs	tail	domestic	catsize
0	1	0	0	1	0	0	1	1	1	1	0	0	4	0	0	1
1	1	0	0	1	0	0	0	1	1	1	0	0	4	1	0	1
2	0	0	1	0	0	1	1	1	1	0	0	1	0	1	0	0
3	1	0	0	1	0	0	1	1	1	1	0	0	4	0	0	1
4	1	0	0	1	0	0	1	1	1	1	0	0	4	1	0	1

```

In [8]:
Y = zoo['type']
Y

```

```

Out[8]:
0      1
1      1
2      4
3      1
4      1
..
96     1
97     6
98     1
99     7
100    2
Name: type, Length: 101, dtype: int64

```

```

In [9]:
# Let's use Grid search CV to find out best value for K

```

```

In [10]:
from sklearn.neighbors import KNeighborsClassifier
from sklearn.model_selection import GridSearchCV

```

```

In [11]:
n_neighbors = np.array(range(1,40))
param_grid = dict(n_neighbors=n_neighbors)

```

```

In [12]:
model = KNeighborsClassifier()
grid = GridSearchCV(estimator=model, param_grid=param_grid)
grid.fit(X, Y)

```

```

Out[12]:
GridSearchCV(estimator=KNeighborsClassifier(),
              param_grid={'n_neighbors': array([ 1,  2,  3,  4,  5,  6,  7,  8,  9, 10, 11, 12, 13, 14, 15, 16, 17, 18, 19, 20, 21, 22, 23, 24, 25, 26, 27, 28, 29, 30, 31, 32, 33, 34, 35, 36, 37, 38, 39])})

```

```

In [13]:
print(grid.best_score_)
print(grid.best_params_)

```

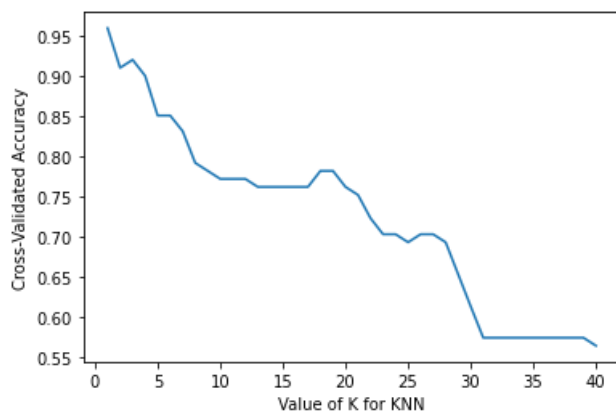
```
0.96
{'n_neighbors': 1}
```

In [14]:

```
# Visualizing the CV results
```

In [15]:

```
import matplotlib.pyplot as plt
%matplotlib inline
# choose k between 1 to 41
k_range = range(1, 41)
k_scores = []
# use iteration to calculator different k in models, then return the average accuracy based on the cross
for k in k_range:
    knn = KNeighborsClassifier(n_neighbors=k)
    scores = cross_val_score(knn, X, Y, cv=5)
    k_scores.append(scores.mean())
# plot to see clearly
plt.plot(k_range, k_scores)
plt.xlabel('Value of K for KNN')
plt.ylabel('Cross-Validated Accuracy')
plt.show()
```



In [16]:

```
#KNN Classification
num_folds = 30
kfold = KFold(n_splits=20)
model = KNeighborsClassifier(n_neighbors=1)
results = cross_val_score(model, X, Y, cv=kfold)
print(results.mean())
```

```
0.9800000000000001
```

In [ ]: