```
In [1]: import pandas as pd
import numpy as np
```

task 01

```
In [2]: #reading data
data = pd.read_csv("Cat_human.csv")
data
```

Out[2]:

	Color	Eye_color	Height	Legs	Moustache	Tail	Weight	label
0	No	black	5.14	2	No	No	70.000000	human
1	No	brown	6.80	2	No	No	64.400000	human
2	Yes	brown	5.00	2	Yes	No	64.800000	human
3	No	blue	5.90	2	No	No	78.800000	human
4	No	blue	6.56	2	No	No	73.200000	human
195	brown	gray	1.14	4	Yes	Yes	2.304511	Cat
196	white	yellow	1.39	4	Yes	Yes	5.687970	Cat
197	white	black	0.53	4	Yes	Yes	6.364662	Cat
198	brown	green	1.03	4	Yes	Yes	6.590226	Cat
199	brown_white	blue	0.83	4	Yes	Yes	7.868421	Cat

200 rows × 8 columns

```
In [3]: #checking if the data is clean or not
data.isna().sum()
```

```
Out[3]: Color
                      0
        Eye_color
                      0
        Height
                      0
        Legs
                      0
        Moustache
                      0
        Tail
                      0
        Weight
                      0
        label
        dtype: int64
```

```
In [5]: #seperatting x and y
x = data.drop("label", axis = 1)
y = data["label"]
```

```
In [6]:
         x , y
Out[6]: (
                      Color Eye_color
                                         Height
                                                  Legs Moustache Tail
                                                                             Weight
                          No
                                  black
                                           5.14
                                                     2
                                                                         70.000000
                                                               No
                         No
                                  brown
                                           6.80
                                                     2
                                                                         64.400000
          1
                                                               No
                                                                     No
          2
                                                     2
                        Yes
                                  brown
                                           5.00
                                                                     No
                                                                         64.800000
                                                              Yes
          3
                                                     2
                         No
                                  blue
                                           5.90
                                                               No
                                                                     No
                                                                         78.800000
          4
                                  blue
                                           6.56
                                                     2
                                                                         73.200000
                         No
                                                               No
                                                                     No
          . .
                         . . .
                                    . . .
                                                              . . .
          195
                      brown
                                           1.14
                                                     4
                                                                          2.304511
                                  gray
                                                              Yes
                                                                    Yes
          196
                      white
                                yellow
                                           1.39
                                                     4
                                                              Yes
                                                                    Yes
                                                                          5.687970
          197
                      white
                                 black
                                           0.53
                                                     4
                                                              Yes
                                                                    Yes
                                                                          6.364662
          198
                      brown
                                 green
                                           1.03
                                                              Yes
                                                                    Yes
                                                                          6.590226
          199
                brown_white
                                  blue
                                           0.83
                                                              Yes
                                                                    Yes
                                                                          7.868421
          [200 rows x 7 columns],
          0
                  human
          1
                  human
          2
                  human
          3
                  human
          4
                  human
          195
                    Cat
          196
                    Cat
          197
                    Cat
          198
                    Cat
          199
                    Cat
          Name: label, Length: 200, dtype: object)
```

task 02

```
In [24]: from sklearn.preprocessing import MinMaxScaler, LabelEncoder
```

```
In [8]: | scaling = MinMaxScaler()
        x scaled = scaling.fit transform(x)
                          <del>comptex data not sapported fill fil . Formatfarray)</del>
             883
                     ) from complex_warning
        File C:\ProgramData\anaconda3\lib\site-packages\sklearn\utils\ array api.p
        y:185, in _asarray_with_order(array, dtype, order, copy, xp)
                     xp, _ = get_namespace(array)
             182
            183 if xp. name in {"numpy", "numpy.array api"}:
                    # Use NumPy API to support order
        --> 185
                     array = numpy.asarray(array, order=order, dtype=dtype)
                     return xp.asarray(array, copy=copy)
            186
            187 else:
        File C:\ProgramData\anaconda3\lib\site-packages\pandas\core\generic.py:207
        0, in NDFrame.__array__(self, dtype)
           2069 def __array__(self, dtype: npt.DTypeLike | None = None) -> np.ndarr
        ay:
        -> 2070
                     return np.asarray(self. values, dtype=dtype)
        ValueError: could not convert string to float: 'No'
```

as MinMaxScaler excpects input as numerical but our features are in cataogorical so we'll use onehot encocoser to make it numerical

```
In [25]: label_encoder = LabelEncoder()

In [30]: c = label_encoder.fit_transform(data["Color"])
    data["Color"] = pd.Series(c)

        e_c = label_encoder.fit_transform(data["Eye_color"])
        data["Eye_color"] = pd.Series(e_c)

        m = label_encoder.fit_transform(data["Moustache"])
        data["Moustache"] = pd.Series(m)

        t = label_encoder.fit_transform(data["Tail"])
```

data["Tail"] = pd.Series(t)

In [31]: data

Out[31]:

	Color	Eye_color	Height	Legs	Moustache	Tail	Weight	label
0	0	0	5.14	2	0	0	70.000000	human
1	0	2	6.80	2	0	0	64.400000	human
2	1	2	5.00	2	1	0	64.800000	human
3	0	1	5.90	2	0	0	78.800000	human
4	0	1	6.56	2	0	0	73.200000	human
195	3	3	1.14	4	1	1	2.304511	Cat
196	6	5	1.39	4	1	1	5.687970	Cat
197	6	0	0.53	4	1	1	6.364662	Cat
198	3	4	1.03	4	1	1	6.590226	Cat
199	4	1	0.83	4	1	1	7.868421	Cat

200 rows × 8 columns

```
In [33]: #seperatting x and y
x = data.drop("label", axis = 1)
y = data["label"]
```

In [34]: x

Out[34]:

	Color	Eye_color	Height	Legs	Moustache	Tail	Weight
0	0	0	5.14	2	0	0	70.000000
1	0	2	6.80	2	0	0	64.400000
2	1	2	5.00	2	1	0	64.800000
3	0	1	5.90	2	0	0	78.800000
4	0	1	6.56	2	0	0	73.200000
195	3	3	1.14	4	1	1	2.304511
196	6	5	1.39	4	1	1	5.687970
197	6	0	0.53	4	1	1	6.364662
198	3	4	1.03	4	1	1	6.590226
199	4	1	0.83	4	1	1	7.868421

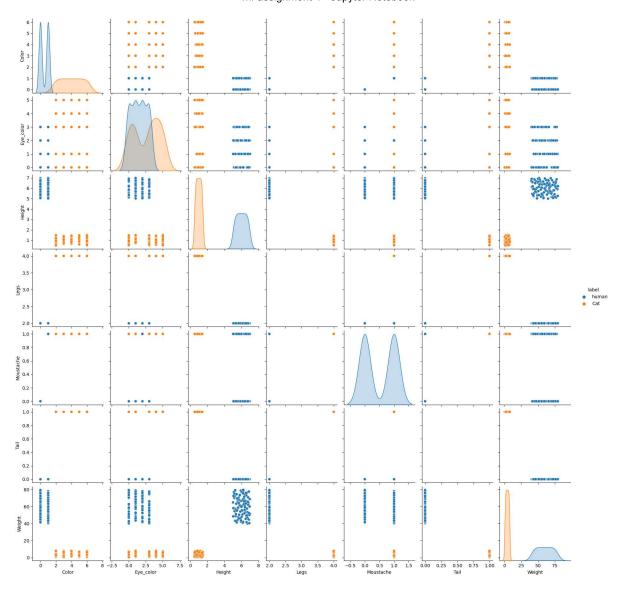
200 rows × 7 columns

```
In [35]: y encoded = label encoder.fit transform(y)
     y_encoded
1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
          0, 01)
In [38]: x scaled = scaling.fit transform(x)
     x_scaled
Out[38]: array([[0.
                 , 0.
                        , 0.71604938, ..., 0.
                                           , 0.
           0.87863464],
                        , 0.97222222, ..., 0.
           0.80783818],
                        , 0.69444444, ..., 1.
          [0.16666667, 0.4
          0.81289507],
          . . . ,
                        , 0.00462963, ..., 1.
                 , 0.
          [1.
                                           , 1.
           0.07414237],
                 , 0.8
                        , 0.08179012, ..., 1.
                                           , 1.
          [0.5
           0.076994 ],
          [0.66666667, 0.2
                        , 0.05092593, ..., 1.
                                           , 1.
           0.09315324]])
In [40]: from sklearn.model selection import train test split
In [41]: np.random.seed(42)
     x_train ,x_test ,y_train ,y_test = train_test_split(x_scaled ,y_encoded ,test_s
     # task 3
In [55]: from sklearn.linear model import LogisticRegression
     import matplotlib.pyplot as plt
In [43]: | model = LogisticRegression()
     model.fit(x_train, y_train)
Out[43]:
     ▼ LogisticRegression
      LogisticRegression()
```

In [52]: import accuracy_score, precision_score, f1_score, recall_score, confusion_matrix

```
In [56]: import seaborn as sns
         # Predict test set
         y_pred = model.predict(x_test)
         #accuracy score
         accuracy = accuracy_score(y_test, y_pred)
         print('Accuracy:', accuracy)
         # Generate confusion matrix
         cm = confusion_matrix(y_test, y_pred)
         print('Confusion Matrix:')
         print(cm)
         #precision score
         precision = precision_score(y_test, y_pred, average='weighted')
         print('Precision:', precision)
         #recall score
         recall = recall_score(y_test, y_pred, average='weighted')
         print('Recall:', recall)
         # Calculate F1 score
         f1 = f1_score(y_test, y_pred, average='weighted')
         print('F1 Score:', f1)
         # Visualize the dataset
         sns.pairplot(data, hue='label')
         plt.show()
```

Accuracy: 1.0
Confusion Matrix:
[[19 0]
 [0 21]]
Precision: 1.0
Recall: 1.0
F1 Score: 1.0



In []: