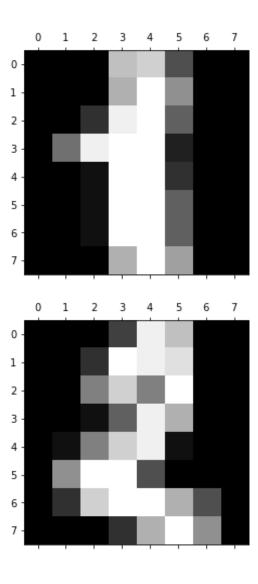
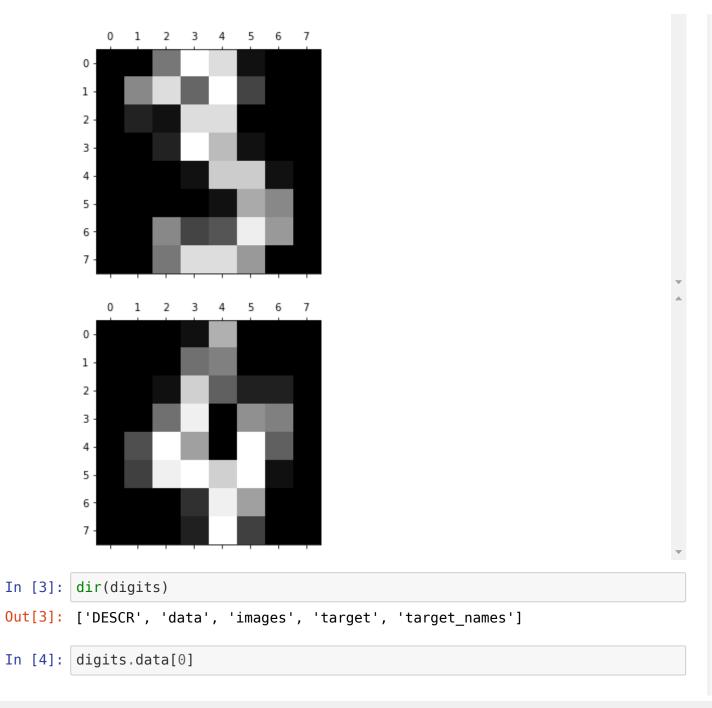
Logistic Regression: Multiclass Classification In [1]: from sklearn.datasets import load digits %matplotlib inline import matplotlib.pyplot as plt digits = load digits() In [2]: plt.gray() for i in range(5): plt.matshow(digits.images[i]) <Figure size 432x288 with 0 Axes> 0 1 2 3 4 5 6 7 1 -2 -3 -5 -





```
Out[4]: array([ 0., 0., 5., 13., 9., 1., 0., 0., 0., 0., 13., 15., 10.,
               15., 5., 0., 0., 3., 15., 2., 0., 11., 8., 0., 0., 4.,
               12., 0., 0., 8., 8., 0., 0., 5., 8., 0., 0., 9., 8.,
                0., 0., 4., 11., 0., 1., 12., 7., 0., 0., 2., 14., 5.,
               10., 12., 0., 0., 0., 6., 13., 10., 0., 0., 0.]
         Create and train logistic regression model
In [5]: from sklearn.linear model import LogisticRegression
         model = LogisticRegression()
In [6]: from sklearn.model selection import train test split
In [15]: X_train, X_test, y_train, y_test = train_test split(digits.data,digits.
         target, test size=0.2)
In [22]: len(X train)
Out[22]: 1437
In [23]: len(X test)
Out[23]: 360
In [24]: model.fit(X train, y train)
         C:\Users\prasa\anaconda3\lib\site-packages\sklearn\linear model\ logist
         ic.py:940: ConvergenceWarning: lbfqs failed to converge (status=1):
         STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
         Increase the number of iterations (max iter) or scale the data as shown
         in:
            https://scikit-learn.org/stable/modules/preprocessing.html
         Please also refer to the documentation for alternative solver options:
            https://scikit-learn.org/stable/modules/linear model.html#logistic-
         regression
           extra warning msg= LOGISTIC SOLVER CONVERGENCE MSG)
```

```
Out[24]: LogisticRegression(C=1.0, class weight=None, dual=False, fit intercept=
         True,
                            intercept scaling=1, l1 ratio=None, max iter=100,
                           multi class='auto', n jobs=None, penalty='l2',
                            random state=None, solver='lbfgs', tol=0.0001, verbo
         se=0,
                           warm start=False)
         Measure accuracy of our model
In [25]: model.score(X test, y test)
Out[25]: 0.966666666666667
In [26]: model.predict(digits.data[0:5])
Out[26]: array([0, 1, 2, 3, 4])
         Confusion Matrix
In [30]: y predicted = model.predict(X test)
In [31]: from sklearn.metrics import confusion matrix
         cm = confusion matrix(y test,y predicted)
         \mathsf{CM}
Out[31]: array([[35, 0,
                         0, 0,
                                                     0],
                                     Θ,
                [ 0, 38,
                         0, 0,
                                 1,
                                     0,
                                                     0],
                                 Θ,
                                         0,
                                             Θ,
                [ 0, 0, 40, 1,
                                     Θ,
                                                     0],
                     0, 1, 30,
                                 0,
                                     0,
                                                 0, 11,
                         0, 0, 32,
                                     0,
                     Θ,
                         Θ,
                             Θ,
                                 Θ,
                                    36,
                                         Ο,
                                                     0],
                             Θ,
                [ 0,
                         Θ,
                                     0, 34, 0,
                                                 0, 0],
                         0, 0,
                                     Θ,
                                         0, 36, 0,
```

```
[0, 0, 0, 0, 0, 2, 0, 0, 31, 0],
                           0, 1, 0, 1, 0, 1, 0, 36]], dtype=int64)
In [33]: import seaborn as sn
         plt.figure(figsize = (10,7))
         sn.heatmap(cm,annot=True)
         plt.xlabel('Predicted')
         plt.ylabel('Truth')
Out[33]: Text(69.0, 0.5, 'Truth')
                            0
                                                                             - 35
                                                        0
                                                                             - 30
                                             0
                                                                             - 25
                                       32
                                             0
          Truth
                                                                             - 20
                                             36
                                                        0
            LO.
                                                                             - 15
            9
                                                        0
                                  0
                                                        36
                                                                             - 10
                                  0
                                             2
                                                        0
                                                             31
                                                                             - 5
                                 1
                                             1
                                                        1
                                                                   36
            9
                      i
                            ż
                                 ż
                                                        7
                                                   6
                                                                   9
                                        Predicted
         Exercise
```

Use sklearn.datasets iris flower dataset to train your model using logistic regression. You need to figure out accuracy of your model and use that to predict different samples in your test dataset. In iris dataset there are 150 samples containing following features,

- 1. Sepal Length
- 2. Sepal Width
- 3. Petal Length
- 4. Petal Width

Using above 4 features you will clasify a flower in one of the three categories,

- 1. Setosa
- 2. Versicolour
- 3. Virginica

