

In [6]: `import pandas as pd`
`ad = pd.read_csv("agridata.csv")`
`ad`

Out[6]:

	outlook	temp	windy	soil	results
0	sunny	NaN	no	NaN	0
1	NaN	cool	yes	NaN	0
2	rain	hot	NaN	NaN	1
3	rain	warm	NaN	NaN	1
4	sunny	NaN	no	NaN	1
5	NaN	cool	yes	NaN	1
6	overcast	NaN	no	NaN	0

In [7]: `# currently we dont have any data of soil.`
`# All 3 external factors like outlook, temperature and windy are depend upon irrigation.`
`# if '0' at perform column means end results of cultivation is bad. so we can take soil as infertile`
`# if '1' at perform column means end results of cultivation is good. so we can take soil as fertile`
`df = pd.read_csv("agridata2.csv")`
`df`

Out[7]:

	outlook	temp	windy	soil	results
0	sunny	NaN	no	infertile	0
1	NaN	cool	yes	infertile	0
2	rain	hot	NaN	fertile	1
3	rain	warm	NaN	fertile	1
4	sunny	NaN	no	fertile	1
5	NaN	cool	yes	fertile	1
6	overcast	NaN	no	infertile	0

In [13]: `# In every column there is NaN which means there is no data and that feild is empty`
`# we can insert some value like "no data" in all empty feilds`

`inputs = df.drop('results',axis='columns')`
`target = df['results']`
`df["outlook"].fillna("No_data", inplace = True)`
`df["temp"].fillna("No_data", inplace = True)`
`df["windy"].fillna("No_data", inplace = True)`
`df`

Out[13]:

	outlook	temp	windy	soil	results
0	sunny	No data	no	infertile	0
1	No data	cool	yes	infertile	0
2	rain	hot	No data	fertile	1
3	rain	warm	No data	fertile	1
4	sunny	No data	no	fertile	1
5	No data	cool	yes	fertile	1
6	overcast	No data	no	infertile	0

In [14]: `# Machine learning algorithms runs on a numbers rather than any given strings`
`# so we need to convert every strings into number using lab encoder algorithm`

`from sklearn.preprocessing import LabelEncoder`
`le_outlook = LabelEncoder()`
`le_temp = LabelEncoder()`
`le_windy = LabelEncoder()`
`le_soil = LabelEncoder()`

`inputs['outlook_n'] = le_outlook.fit_transform(inputs['outlook'])`
`inputs['temp_n'] = le_temp.fit_transform(inputs['temp'])`
`inputs['windy_n'] = le_windy.fit_transform(inputs['windy'])`
`inputs['soil_n'] = le_soil.fit_transform(inputs['soil'])`
`inputs`

Out[14]:

	outlook	temp	windy	soil	outlook_n	temp_n	windy_n	soil_n
0	sunny	No data	no	infertile	3	0	1	1
1	No data	cool	yes	infertile	0	1	2	1
2	rain	hot	No data	fertile	2	2	0	0
3	rain	warm	No data	fertile	2	3	0	0
4	sunny	No data	no	fertile	3	0	1	0
5	No data	cool	yes	fertile	0	1	2	0
6	overcast	No data	no	infertile	1	0	1	1

In [15]: `# New encoded columns has been added which represents each string as an unique number on a column`
`# Now we can remove string values columns as now there is no need to store`

`inputs_n = inputs.drop(['outlook', 'temp', 'windy', 'soil'],axis='columns')`
`inputs_n`

Out[15]:

	outlook_n	temp_n	windy_n	soil_n
0	3	0	1	1
1	0	1	2	1
2	2	2	0	0
3	2	3	0	0
4	3	0	1	0
5	0	1	2	0
6	1	0	1	1

In [16]: `# Our target data is results cloumn in which 0 represents bad results and 1 respresents good results of cultivation.`

`target`

Out[16]:

0	0
1	0
2	1
3	1
4	1
5	1
6	0

Name: results, dtype: int64

In [17]: `#Now we are adding Decision tree algorithm to predict from the labeled or given data`
`#This label data model helps machine learning to train or learn the model so that it can predict acc urately based on the given sets`

`from sklearn import tree`
`model = tree.DecisionTreeClassifier()`
`model.fit(inputs_n, target)`

Out[17]: DecisionTreeClassifier(class_weight=None, criterion='gini', max_depth=None, max_features=None, max_leaf_nodes=None, min_impurity_decrease=0.0, min_impurity_split=None, min_samples_leaf=1, min_samples_split=2, min_weight_fraction_leaf=0.0, presort=False, random_state=None, splitter='best')

In [18]: `# Now we can predict our score which will be going to 1`
`# For complex data set the value score is always less than 1`

`model.score(inputs_n,target)`

Out[18]: 1.0

In [19]: `#Now its time to test or predict our data model`
`# we have to test by assigning our 4 columns value in an array of predict function`
`# 4 columns are outlook, temperature, windy and soil respectively`
`# In outlook it is 2 which represents 'rain'`
`# In Temperature it is 1 which represents 'cool'`
`# In windy it is 2 which respresents 'yes' means it is windy`
`# In soil it is 1 which represents infertile`
`# so our target results will be 0 which means end result of cultivation is not good on that day`
`model.predict([[2,1,1,1]])`

Out[19]: array([0], dtype=int64)

In [20]: `# we can follow the same method by taking another data values`
`# For better understanding, just take a look at output row [14] to take data values.`
`model.predict([[2,0,0,0]])`

Out[20]: array([1], dtype=int64)

In [21]: `model.predict([[2,1,1,0]])`

Out[21]: array([1], dtype=int64)

In [22]: `model.predict([[0,0,0,0]])`

Out[22]: array([1], dtype=int64)

In []: