

```
In [1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
```

```
In [2]: salary_train = pd.read_csv('SalaryData_Train(1).csv')
salary_train
```

Out[2]:

	age	workclass	education	educationno	maritalstatus	occupation	relationship	race	sex
0	39	State-gov	Bachelors	13	Never-married	Adm-clerical	Not-in-family	White	M
1	50	Self-emp-not-inc	Bachelors	13	Married-civ-spouse	Exec-managerial	Husband	White	M
2	38	Private	HS-grad	9	Divorced	Handlers-cleaners	Not-in-family	White	M
3	53	Private	11th	7	Married-civ-spouse	Handlers-cleaners	Husband	Black	M
4	28	Private	Bachelors	13	Married-civ-spouse	Prof-specialty	Wife	Black	F
...
30156	27	Private	Assoc-acdm	12	Married-civ-spouse	Tech-support	Wife	White	F
30157	40	Private	HS-grad	9	Married-civ-spouse	Machine-op-inspct	Husband	White	M
30158	58	Private	HS-grad	9	Widowed	Adm-clerical	Unmarried	White	F
30159	22	Private	HS-grad	9	Never-married	Adm-clerical	Own-child	White	M
30160	52	Self-emp-inc	HS-grad	9	Married-civ-spouse	Exec-managerial	Wife	White	F

30161 rows × 10 columns

```
In [3]: salary_test = pd.read_csv('SalaryData_Test(1).csv')
salary_train
```

Out[3]:

	age	workclass	education	educationno	maritalstatus	occupation	relationship	race	sex
0	39	State-gov	Bachelors	13	Never-married	Adm-clerical	Not-in-family	White	M
1	50	Self-emp-not-inc	Bachelors	13	Married-civ-spouse	Exec-managerial	Husband	White	M
2	38	Private	HS-grad	9	Divorced	Handlers-cleaners	Not-in-family	White	M
3	53	Private	11th	7	Married-civ-spouse	Handlers-cleaners	Husband	Black	M
4	28	Private	Bachelors	13	Married-civ-spouse	Prof-specialty	Wife	Black	F
...
30156	27	Private	Assoc-acdm	12	Married-civ-spouse	Tech-support	Wife	White	F
30157	40	Private	HS-grad	9	Married-civ-spouse	Machine-op-inspct	Husband	White	M
30158	58	Private	HS-grad	9	Widowed	Adm-clerical	Unmarried	White	F
30159	22	Private	HS-grad	9	Never-married	Adm-clerical	Own-child	White	M
30160	52	Self-emp-inc	HS-grad	9	Married-civ-spouse	Exec-managerial	Wife	White	F

30161 rows × 10 columns



```
In [4]: #initial analysis
salary_train.shape
```

Out[4]: (30161, 10)

```
In [5]: salary_train.isna().sum()
```

```
Out[5]: age                0
workclass              0
education              0
educationno            0
maritalstatus          0
occupation             0
relationship           0
race                   0
sex                    0
capitalgain            0
capitalloss            0
hoursperweek           0
native                 0
Salary                 0
dtype: int64
```

```
In [6]: salary_train.dtypes
```

```
Out[6]: age                int64
workclass              object
education              object
educationno            int64
maritalstatus          object
occupation             object
relationship           object
race                   object
sex                    object
capitalgain            int64
capitalloss            int64
hoursperweek           int64
native                 object
Salary                 object
dtype: object
```

```
In [7]: # converting object data type to int data type.
from sklearn.preprocessing import LabelEncoder
le= LabelEncoder()
```

```
In [8]: salary_train['workclass'] = le.fit_transform(salary_train['workclass'])
salary_train['education'] = le.fit_transform(salary_train['education'])
salary_train['maritalstatus'] = le.fit_transform(salary_train['maritalstatus'])
salary_train['occupation'] = le.fit_transform(salary_train['occupation'])
salary_train['relationship'] = le.fit_transform(salary_train['relationship'])
salary_train['race'] = le.fit_transform(salary_train['race'])
salary_train['sex'] = le.fit_transform(salary_train['sex'])
salary_train['native'] = le.fit_transform(salary_train['native'])
salary_train['Salary'] = le.fit_transform(salary_train['Salary'])
salary_train
```

Out[8]:

	age	workclass	education	educationno	maritalstatus	occupation	relationship	race	sex
0	39	5	9	13	4	0	1	4	1
1	50	4	9	13	2	3	0	4	1
2	38	2	11	9	0	5	1	4	1
3	53	2	1	7	2	5	0	2	1
4	28	2	9	13	2	9	5	2	0
...
30156	27	2	7	12	2	12	5	4	0
30157	40	2	11	9	2	6	0	4	1
30158	58	2	11	9	6	0	4	4	0
30159	22	2	11	9	4	0	3	4	1
30160	52	3	11	9	2	3	5	4	0

30161 rows × 14 columns

```
In [9]: salary_train.dtypes
```

```
Out[9]: age                int64
workclass                int32
education                int32
educationno             int64
maritalstatus           int32
occupation              int32
relationship            int32
race                   int32
sex                   int32
capitalgain            int64
capitalloss            int64
hoursperweek           int64
native                 int32
Salary                 int32
dtype: object
```

Splitting the Dataset

```
In [10]: X = salary_train.drop(['Salary'], axis=1)
y = salary_train['Salary']
```

```
In [11]: X
```

```
Out[11]:
```

	age	workclass	education	educationno	maritalstatus	occupation	relationship	race	sex
0	39	5	9	13	4	0	1	4	1
1	50	4	9	13	2	3	0	4	1
2	38	2	11	9	0	5	1	4	1
3	53	2	1	7	2	5	0	2	1
4	28	2	9	13	2	9	5	2	0
...
30156	27	2	7	12	2	12	5	4	0
30157	40	2	11	9	2	6	0	4	1
30158	58	2	11	9	6	0	4	4	0
30159	22	2	11	9	4	0	3	4	1
30160	52	3	11	9	2	3	5	4	0

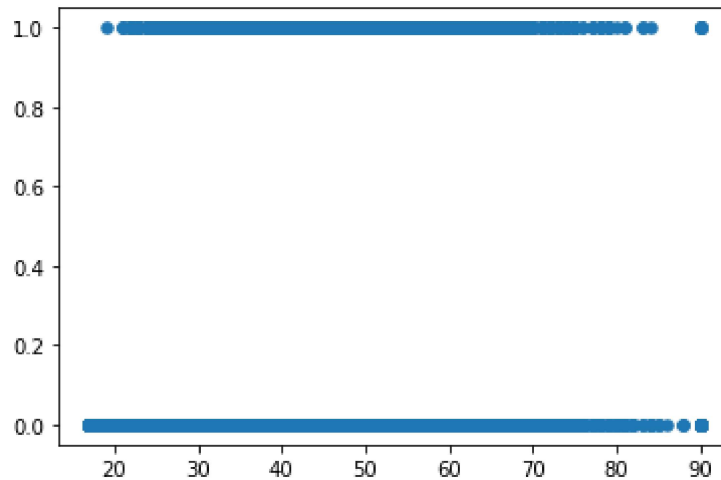
30161 rows × 13 columns



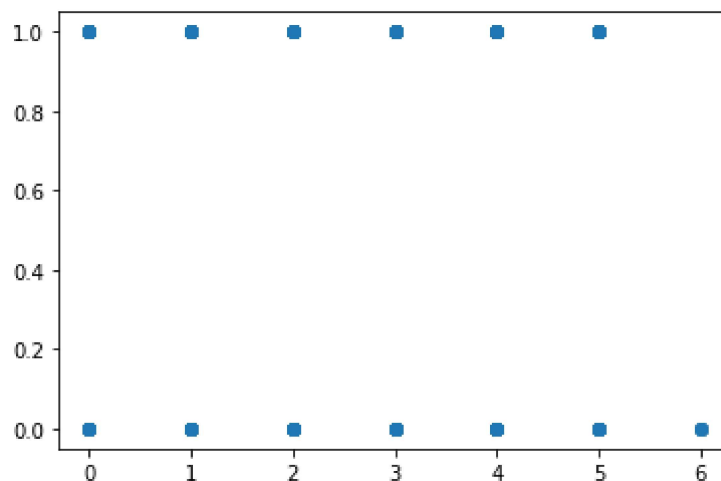
```
In [12]: y
```

```
Out[12]: 0      0
1      0
2      0
3      0
4      0
..
30156   0
30157   1
30158   0
30159   0
30160   1
Name: Salary, Length: 30161, dtype: int32
```

```
In [13]: plt.scatter(salary_train['age'], y, s=30, alpha=1)
plt.show()
```



```
In [14]: plt.scatter(salary_train['workclass'], y, s=30, alpha=1)
plt.show()
```



Model Building

```
In [15]: from sklearn.model_selection import train_test_split
from sklearn.metrics import confusion_matrix, plot_confusion_matrix, accuracy_score
```

```
In [16]: X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.20, random_
```



```
In [17]: X_train.shape, y_train.shape, X_test.shape, y_test.shape
```

```
Out[17]: ((24128, 13), (24128,), (6033, 13), (6033,))
```

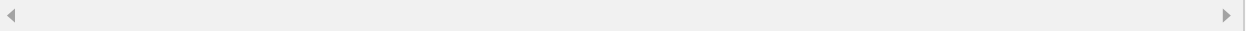
```
In [18]: from sklearn.svm import SVC
```

```
In [ ]: classifier = SVC(kernel='linear', C= 1, gamma = 1)
classifier.fit(X_train, y_train)
y_test_pred = classifier.predict(X_test)
```

```
In [ ]: accuracy_score(y_test, y_test_pred)
```

```
In [ ]: confusion_matrix(y_test, y_test_pred)
```

```
In [ ]: plot_confusion_matrix(classifier, X_test, y_test, cmap='plasma')
plt.show()
```



```
In [ ]:
```

