

```
Import libraries

In [1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import matplotlib inline
import seaborn as sns
sns.set(color_codes=True)
from sklearn.preprocessing import LabelEncoder

Importing the dataset

In [2]: dataset = pd.read_csv("Social_Network_Ads.csv")

In [3]: dataset.head()

Out[3]:   User ID  Gender  Age  EstimatedSalary  Purchased
0  15624510    Male   19           19000           0
1  15810944    Male   35           20000           0
2  15668575    Female  26           43000           0
3  15603246    Female  27           57000           0
4  15804002    Male   19           76000           0

In [4]: dataset.shape

Out[4]: (400, 5)

In [5]: dataset.columns

Out[5]: Index(['User ID', 'Gender', 'Age', 'EstimatedSalary', 'Purchased'], dtype='object')

In [6]: dataset.isna().any()

Out[6]: User ID      False
Gender      False
Age         False
EstimatedSalary  False
Purchased    False
dtype: bool

In [7]: dataset.isna().sum()

Out[7]: User ID      0
Gender      0
Age         0
EstimatedSalary  0
Purchased    0
dtype: int64

In [8]: dataset["Gender"].unique()

Out[8]: array(['Male', 'Female'], dtype=object)

In [9]: dataset["Gender"].replace({'Male':'0', 'Female':'1'},inplace=True)

In [10]: dataset

Out[10]:   User ID  Gender  Age  EstimatedSalary  Purchased
0  15624510      0   19           19000           0
1  15810944      0   35           20000           0
2  15668575      1   26           43000           0
3  15603246      1   27           57000           0
4  15804002      0   19           76000           0
...
395 15691863      1   46           41000           1
396 15706071      0   51           23000           1
397 15654296      1   50           20000           1
398 15755018      0   36           33000           0
399 15594041      1   49           36000           1
400 rows x 5 columns

In [21]: x=dataset.iloc[:,0:4]

In [22]: y=dataset.iloc[:,4]

Splitting the dataset into the Training set and Test set

In [23]: from sklearn.model_selection import train_test_split

In [25]: X_train, X_test, y_train, y_test = train_test_split(x, y, test_size=0.2)

In [26]: len(X_train)

Out[26]: 320

In [27]: len(X_test)

Out[27]: 80

Feature Scaling

In [28]: from sklearn.preprocessing import StandardScaler

In [29]: sc_X = StandardScaler()

In [30]: X_train = sc_X.fit_transform(X_train)

In [31]: X_test = sc_X.transform(X_test)

Fitting SVM to the Training set

In [32]: from sklearn.svm import SVC

In [33]: model=SVC(kernel='linear')

In [34]: model.fit(X_train,y_train)

Out[34]: SVC(kernel='linear')

Predicting the Test set results

In [35]: model.predict(X_test)

Out[35]: array([0, 0, 0, 0, 1, 1, 0, 0, 1, 0, 0, 0, 0, 0, 1, 0, 1, 0, 0, 1, 0, 0, 1, 0, 0, 0,
        0, 0, 1, 1, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
        0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 1, 0, 0, 0, 0, 0,
        0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0], dtype=int64)

In [36]: X_test

Out[36]: array([[ 0.24944539,  1.01892912, -0.31388835, -1.3079089 ],
        [-1.08709592,  1.01892912, -0.31388835,  0.01947796],
        [-1.3221496 ,  1.01892912, -1.06981891,  0.39460903],
        [ 1.66898307,  1.01892912, -0.21939703,  0.1349029 ],
        [ 1.2342395 ,  1.01892912,  1.10348145,  0.51003397],
        [ 1.2196318 ,  1.01892912,  2.04839465,  1.0871587 ],
        [ 1.32416858,  1.01892912, -0.88083627,  0.36575279],
        [-0.76200532,  1.01892912, -0.31388835, -0.586503 ],
        [-0.16609669, -0.98142253, -0.21939703,  2.1259832 ],
        [-0.53431431,  1.01892912,  1.10348145, -1.45219009],
        [ 0.39910776, -0.98142253, -1.16431023,  0.4523215 ],
        [-1.02743265,  1.01892912, -0.21939703, -0.586503 ],
        [ 0.75595295,  1.01892912, -1.35329288,  0.53889021],
        [ 1.43944754,  1.01892912, -0.21939703,  1.6057095],
        [-1.00279614,  1.01892912,  1.4477842 , -1.2213182 ],
        [-1.34346592, -0.98142253,  1.00899013,  0.53889021],
        [-0.99947775,  1.01892912,  0.91449881, -1.15246396],
        [ 0.4607016 , -0.98142253, -0.12490571,  0.0483342 ],
        [-1.37821347,  1.01892912,  1.00899013, -0.99049031],
        [-0.91352622, -0.98142253,  0.06407693,  0.01947796],
        [ 0.45393915, -0.98142253, -1.06981891,  0.56774645],
        [ 1.61710683, -0.98142253, -1.82574948,  0.42346527],
        [-0.45162606,  1.01892912, -0.12490571,  0.01947796],
        [-0.52273179,  1.01892912, -1.16431023, -1.53875879],
        [-1.36277925, -0.98142253,  2.04839465,  0.91402128],
        [ 0.46704856,  1.01892912, -0.21939703, -0.586503 ],
        [ 0.10045528, -0.98142253, -0.21939703,  0.82745257],
        [-1.36811553,  1.01892912,  0.06407693,  0.1349029 ],
        [-1.55808563, -0.98142253, -0.69185363,  0.10604667],
        [-1.15273952,  1.01892912, -0.69185363, -1.59647127],
        [ 1.05369508,  1.01892912, -0.31388835,  0.0483342 ],
        [-0.64922913,  1.01892912,  0.72551617, -1.10591525],
        [ 1.2616622 ,  1.01892912, -0.50287099, -1.23143888],
        [ 0.00521813,  1.01892912,  1.76492069, -1.27905267],
        [-1.24827078, -0.98142253, -0.50287099, -0.78849665],
        [-1.05478532, -0.98142253, -0.31388835, -0.32679687],
        [-0.03775063, -0.98142253, -0.97532759,  0.53889021],
        [ 0.02729074,  1.01892912, -1.35329288, -1.10591525],
        [ 1.4057504 , -0.98142253,  1.00899013, -1.22131402 ],
        [-1.51120934,  1.01892912, -0.59736231,  0.4523215 ],
        [ 1.50719196,  1.01892912, -0.97532759, -0.44222182],
        [ 1.71384257, -0.98142253, -0.69185363, -0.12480322],
        [-1.63517169, -0.98142253, -0.31388835, -0.07719043],
        [-0.78456392,  1.01892912, -0.78634495, -1.31800891],
        [ 0.58423197,  1.01892912, -0.12490571,  0.65431516],
        [-1.15978427,  1.01892912,  0.25305957,  0.0483342 ],
        [ 0.69687791, -0.98142253,  0.63204845, -1.10591525],
        [ 0.44755266,  1.01892912, -1.73125816,  0.33689656],
        [ 0.17740129, -0.98142253,  0.72551617, -1.36562138],
        [-1.36158898,  1.01892912, -1.73125816, -1.36562138],
        [-1.24829316,  1.01892912, -1.06981891, -0.38480351],
        [ 0.64427339,  1.01892912, -0.78634495,  0.25032785],
        [-1.0229509 ,  1.01892912, -0.03041439,  1.20258364],
        [-0.24171549, -0.98142253, -0.69185363,  0.35565311],
        [ 0.2210283 , -0.98142253, -0.31388835, -1.39447761],
        [ 0.7519894 , -0.98142253, -1.73125816,  0.4523215 ],
        [-0.28963888,  1.01892912, -1.16431023,  0.0483342 ],
        [-0.4137973 , -0.98142253, -1.82574948, -0.00378282],
        [-1.47688195,  1.01892912,  0.25305957,  0.0483342 ],
        [-0.00894139,  1.01892912,  0.15856825,  0.0483342 ],
        [-1.62781882, -0.98142253,  0.25305957, -0.73078418],
        [ 0.41180064, -0.98142253, -0.03041439, -0.2690844 ],
        [ 0.44417734,  1.01892912,  1.10348145, -0.75964042],
        [-0.69413065, -0.98142253, -1.54227552, -1.50990256],
        [-1.55690917, -0.98142253, -0.12490571,  0.10604667],
        [ 0.76036447, -0.98142253, -1.9202408 , -0.52879053],
        [ 0.61318126, -0.98142253, -0.03041439, -0.32679687],
        [ 0.1756366 , -0.98142253, -0.59736231, -1.50990256],
        [ 0.76465034, -0.98142253,  0.72551617, -1.34864682],
        [ 1.00800926,  1.01892912, -1.06981891, -1.53875879],
        [ 0.0300775 , -0.98142253, -1.35329288, -0.44222182],
        [-1.64143213,  1.01892912,  0.91449881,  1.75085213],
        [ 1.4875563 ,  1.01892912, -0.78634495, -0.24022817],
        [-0.67436894,  1.01892912, -0.40837967,  0.0483342 ],
        [ 0.51270607,  1.01892912, -1.06981891,  1.92398955],
        [ 1.52855869, -0.98142253, -1.63676684,  0.0483342 ],
        [ 1.42076425,  1.01892912, -0.12490571,  0.27918409],
        [ 0.9158253 , -0.98142253, -0.31388835, -0.93277783],
        [-0.16378579,  1.01892912,  0.44204221,  1.8085646 ],
        [ 0.30063535, -0.98142253, -0.12490571,  0.22147161]])

In [37]: model.score(X_test,y_test)

Out[37]: 0.8

In [45]: y_pred = model.predict(X_test)
y_pred

Out[45]: array([0, 0, 0, 0, 1, 1, 0, 0, 1, 0, 0, 0, 0, 0, 1, 0, 1, 0, 0, 1, 0, 0, 1, 0, 0, 0,
        0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
        0, 1, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 1, 0, 0, 0, 1, 0, 0, 0, 1, 0, 0, 0,
        0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 1, 0], dtype=int64)

Making the Confusion Matrix

In [46]: from sklearn.metrics import confusion_matrix
cm = confusion_matrix(y_test, y_pred)

In [47]: print(cm)

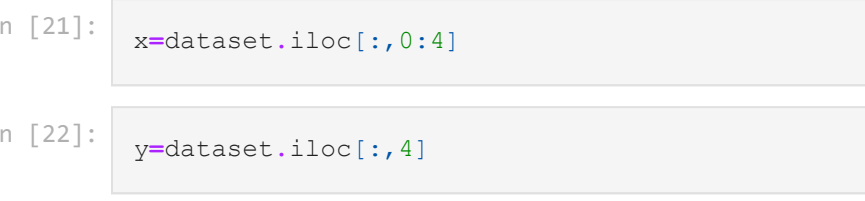
[[54  5]
 [11 10]]

Visualising the Training set results

In [56]: X = dataset.iloc[:, [2, 3]].values
y = dataset.iloc[:,4].values
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.25, random_state = 0)
from sklearn.preprocessing import StandardScaler
sc= StandardScaler()
X_train = sc.fit_transform(X_train)
X_test = sc.transform(X_test)
from sklearn.svm import SVC
classifier = SVC(kernel = 'linear', random_state = 0)
classifier.fit(X_train, y_train)
y_pred = classifier.predict(X_test)
from sklearn.metrics import confusion_matrix
cm = confusion_matrix(y_test, y_pred)
from matplotlib.colors import ListedColormap
X_set, y_set = X_train, y_train
XI, X2 = np.meshgrid(np.arange(start = X_set[:, 0].min() - 1, stop = X_set[:, 0].max() + 1, step = 0.01),
                    np.arange(start = X_set[:, 1].min() - 1, stop = X_set[:, 1].max() + 1, step = 0.01))
plt.contourf(XI, X2, classifier.predict(np.array([X1.ravel(), X2.ravel()]).T).reshape(XI.shape),
             alpha = 0.75, cmap = ListedColormap(('red','green')))
plt.xlim(X1.min(), X1.max())
plt.ylim(X2.min(), X2.max())
for i, j in enumerate(np.unique(y_set)):
    plt.scatter(X_set[y_set == j, 0], X_set[y_set == j, 1],
               c = ListedColormap(('red','green'))(i), label = j)
plt.title('SVM(Training set)')
plt.xlabel('Age')
plt.ylabel('Estimated Salary')
plt.legend()
plt.show()

/*c* argument looks like a single numeric RGB or RGBA sequence, which should be avoided as value-mapping will have precedence in case its length matches with '*x' & '*y'. Please use the *color* keyword-argument or provide a 2D array with a single row if you intend to specify the same RGB or RGBA value for all points.
*c* argument looks like a single numeric RGB or RGBA sequence, which should be avoided as value-mapping will have precedence in case its length matches with '*x' & '*y'. Please use the *color* keyword-argument or provide a 2D array with a single row if you intend to specify the same RGB or RGBA value for all points.
```

SVM(Training set)



Visualising the Test set results

```
In [57]: X = dataset.iloc[:, [2, 3]].values
y = dataset.iloc[:,4].values
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.25, random_state = 0)
from sklearn.preprocessing import StandardScaler
sc= StandardScaler()
X_train = sc.fit_transform(X_train)
X_test = sc.transform(X_test)
from sklearn.svm import SVC
classifier = SVC(kernel = 'linear', random_state = 0)
classifier.fit(X_train, y_train)
y_pred = classifier.predict(X_test)
from sklearn.metrics import confusion_matrix
cm = confusion_matrix(y_test, y_pred)
from matplotlib.colors import ListedColormap
X_set, y_set = X_test, y_test
XI, X2 = np.meshgrid(np.arange(start = X_set[:, 0].min() - 1, stop = X_set[:, 0].max() + 1, step = 0.01),
                    np.arange(start = X_set[:, 1].min() - 1, stop = X_set[:, 1].max() + 1, step = 0.01))
plt.contourf(XI, X2, classifier.predict(np.array([X1.ravel(), X2.ravel()]).T).reshape(XI.shape),
             alpha = 0.75, cmap = ListedColormap(('red','green')))
plt.xlim(X1.min(), X1.max())
plt.ylim(X2.min(), X2.max())
for i, j in enumerate(np.unique(y_set)):
    plt.scatter(X_set[y_set == j, 0], X_set[y_set == j, 1],
               c = ListedColormap(('red','green'))(i), label = j)
plt.title('SVM(Test set)')
plt.xlabel('Age')
plt.ylabel('Estimated Salary')
plt.legend()
plt.show()

/*c* argument looks like a single numeric RGB or RGBA sequence, which should be avoided as value-mapping will have precedence in case its length matches with '*x' & '*y'. Please use the *color* keyword-argument or provide a 2D array with a single row if you intend to specify the same RGB or RGBA value for all points.
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```

