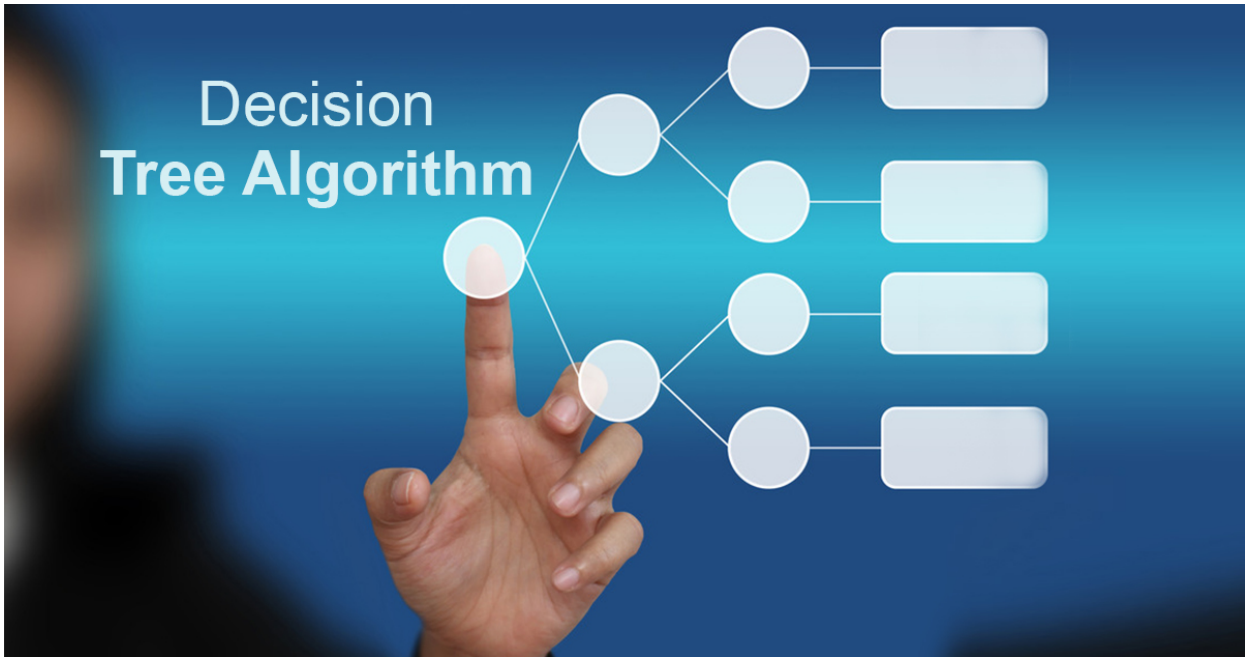
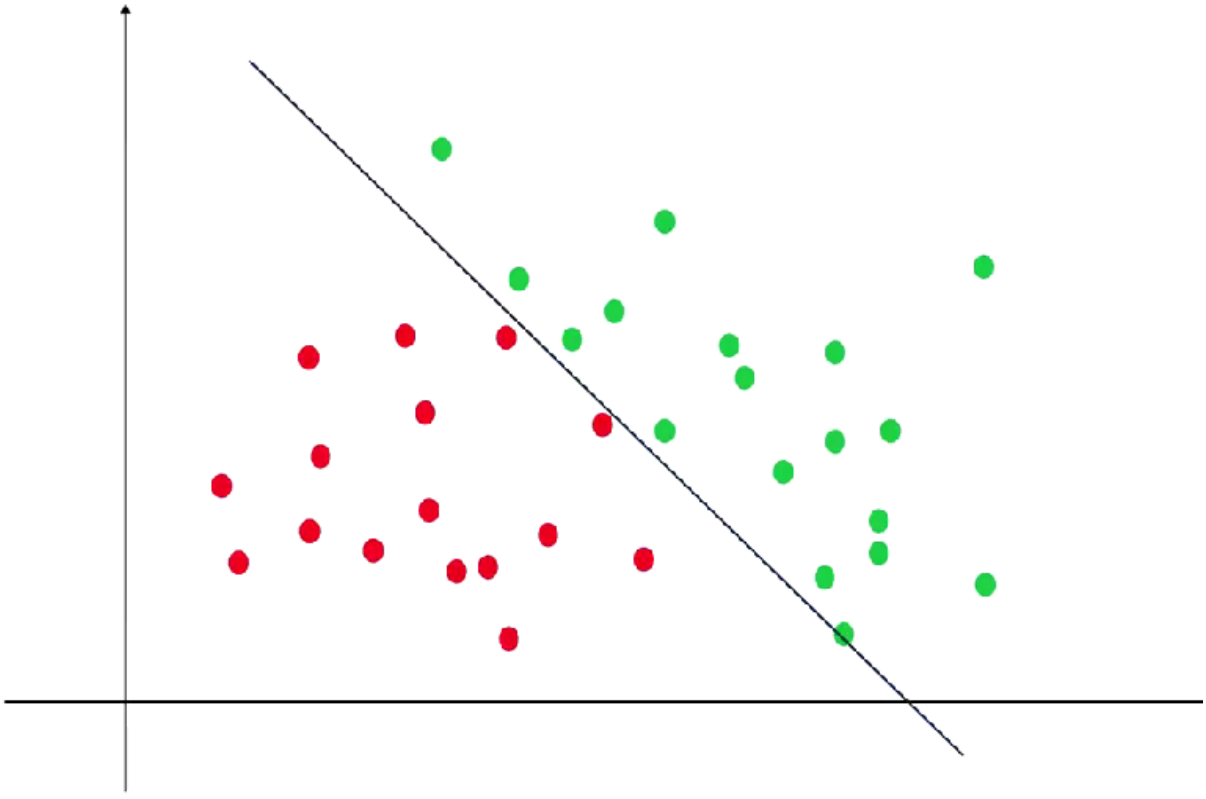
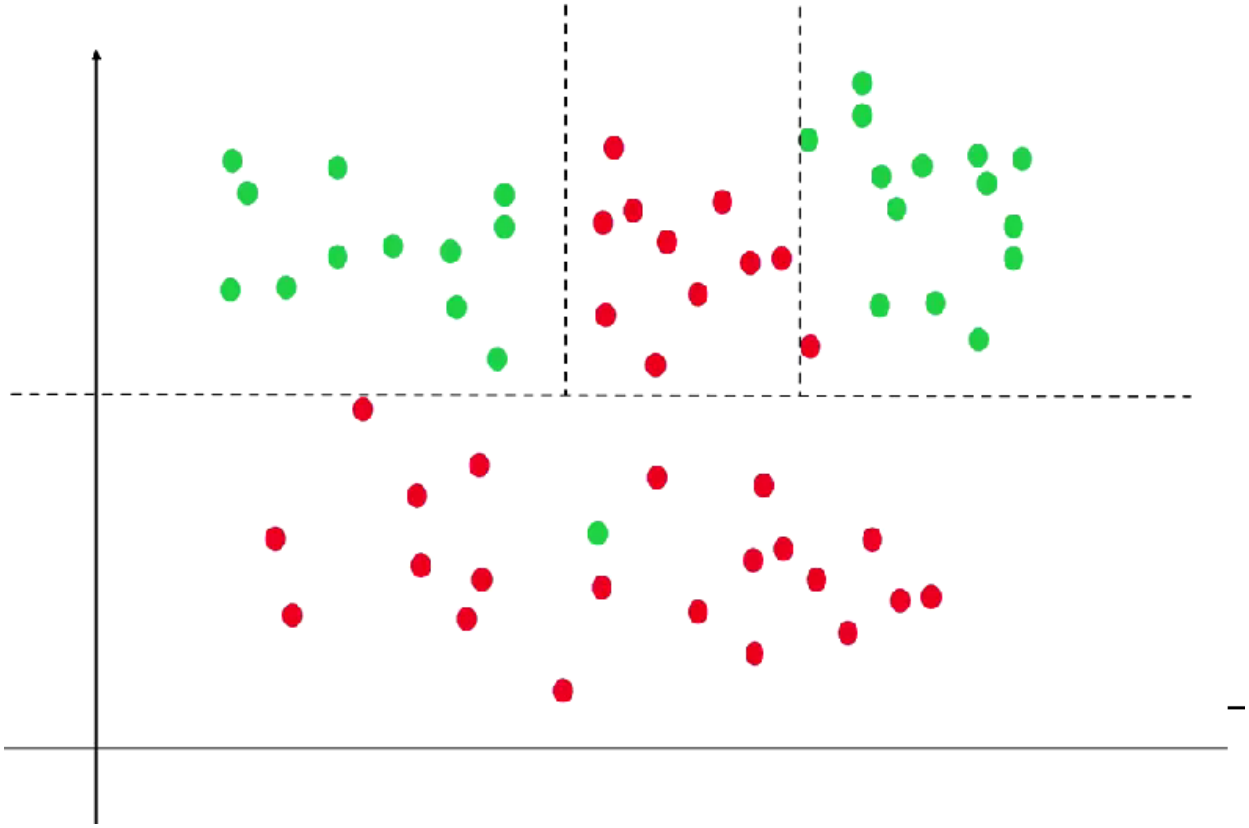
Classification Problem:Decision Tree



When you have a dataset like this it's easier to draw a decision boundry using LogisticRegression.

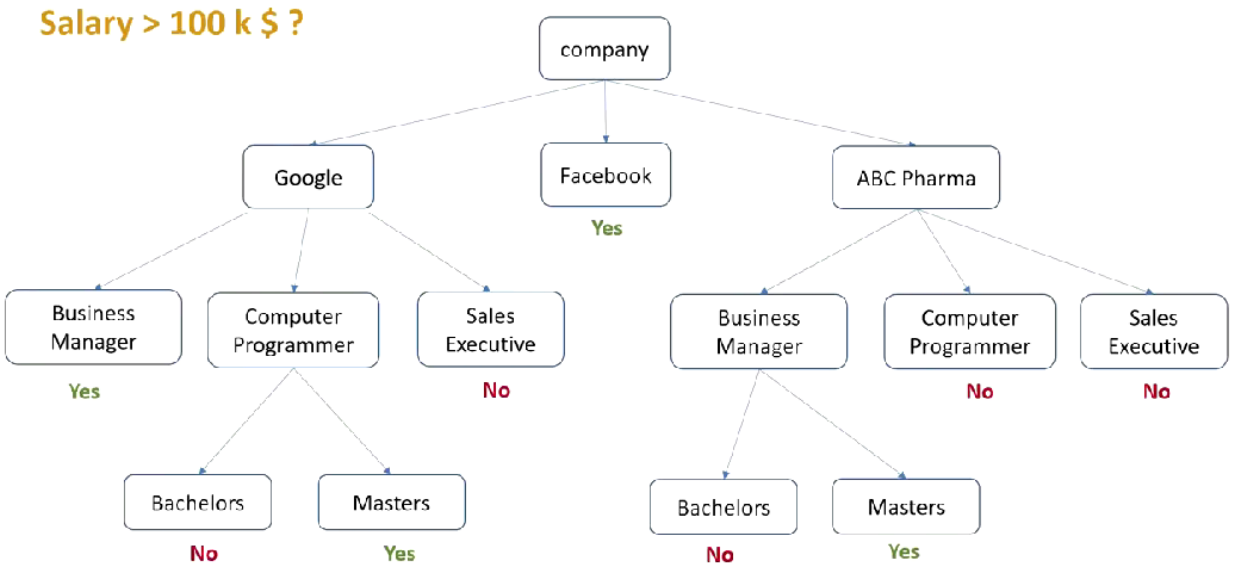


But if your dataset is complicate like this you cannot just draw a single line. You might have to split your dataset again and again to come up with the decision boundaries and this is what decision algorithm does for you.

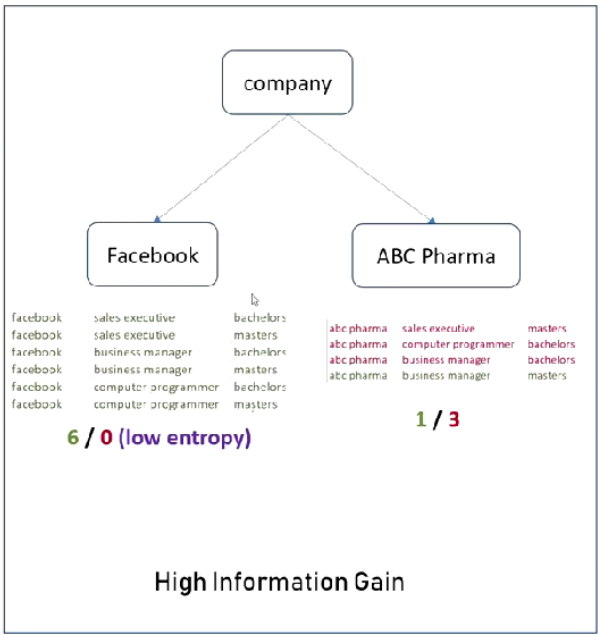


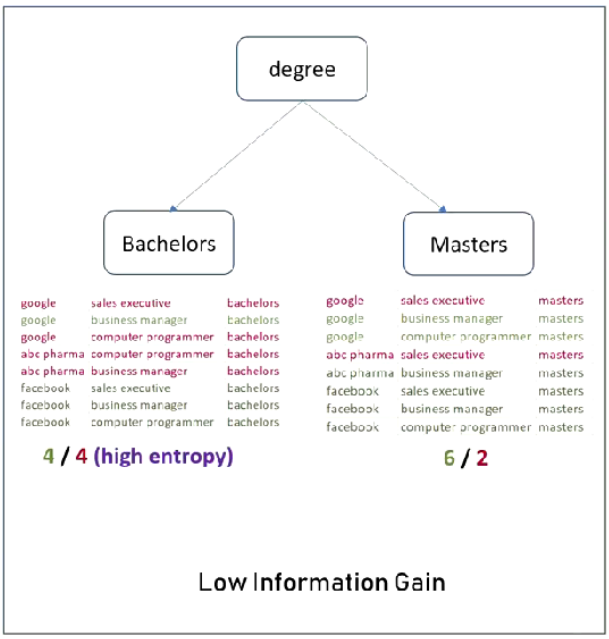
Problem: We will use this particular dataset where you try to predict if person salaries more than 100K$ based on the company and the job title and the degree that he has.

Company	Job	Degree	Salary_more_then_100k
google	sales executive	bachelors	0
google	sales executive	masters	0
google	business manager	bachelors	1
google	business manager	masters	1
google	computer programmer	bachelors	0
google	computer programmer	masters	1
abc pharma	sales executive	masters	0
abc pharma	computer programmer	bachelors	0
abc pharma	business manager	bachelors	0
abc pharma	business manager	masters	1
facebook	sales executive	bachelors	1
facebook	sales executive	masters	1
facebook	business manager	bachelors	1
facebook	business manager	masters	1
facebook	computer programmer	bachelors	1
facebook	computer programmer	masters	1



How do you select ordering of features?





entropy is the measure of randomness in your samples. low entropy means there is no randomness. low entropy => High information gain => better choice for ordering features.

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

In [3]:
```
df = pd.read_csv("D:/Data_Science/My Github/Machine-Learning-with-Python/9. Decisi
df.head()
```

Out[3]:

	company	job	degree	salary_more_then_100k
0	google	sales executive	bachelors	0
1	google	sales executive	masters	0
2	google	business manager	bachelors	1
3	google	business manager	masters	1
4	google	computer programmer	bachelors	0

In [4]:
```
# divided my dataframe between independant variable(inputs) & target varibable
inputs = df.drop('salary_more_then_100k',axis='columns')
```

In [5]:
```
target = df['salary_more_then_100k']
```

ML algorithm can only work on numbers, so we have to converts input columns into numbers. One way to do this in to use LabelEncoder.

In [6]:
```
from sklearn.preprocessing import LabelEncoder
```

I have 3 columns so I have to create 3 different objects.

In [7]:
```
le_company = LabelEncoder()
le_job = LabelEncoder()
le_degree = LabelEncoder()
```

Once you have those 3, in your inputs dataframe you are creating 1 more column for each by using fit_transform method to convert them into numbers.

In [8]:
```
inputs['company_n'] = le_company.fit_transform(inputs['company'])
inputs['job_n'] = le_job.fit_transform(inputs['job'])
inputs['degree_n'] = le_degree.fit_transform(inputs['degree'])
```

In [9]:
```
inputs
```

Out[9]:

	company	job	degree	company_n	job_n	degree_n
0	google	sales executive	bachelors	2	2	0
1	google	sales executive	masters	2	2	1
2	google	business manager	bachelors	2	0	0
3	google	business manager	masters	2	0	1
4	google	computer programmer	bachelors	2	1	0
5	google	computer programmer	masters	2	1	1
6	abc pharma	sales executive	masters	0	2	1
7	abc pharma	computer programmer	bachelors	0	1	0
8	abc pharma	business manager	bachelors	0	0	0
9	abc pharma	business manager	masters	0	0	1
10	facebook	sales executive	bachelors	1	2	0
11	facebook	sales executive	masters	1	2	1
12	facebook	business manager	bachelors	1	0	0
13	facebook	business manager	masters	1	0	1
14	facebook	computer programmer	bachelors	1	1	0
15	facebook	computer programmer	masters	1	1	1

Next step is to drop the real columns for just having numbers.

In [10]:
```
inputs_n = inputs.drop(['company','job','degree'],axis='columns')
```

In [16]:
```
inputs_n
```

Out[16]:

	company_n	job_n	degree_n
0	2	2	0
1	2	2	1
2	2	0	0
3	2	0	1
4	2	1	0
5	2	1	1
6	0	2	1
7	0	1	0
8	0	0	0
9	0	0	1
10	1	2	0
11	1	2	1
12	1	0	0
13	1	0	1
14	1	1	0
15	1	1	1

In [17]:
```
target
```

Out[17]:
```
0     0
1     0
2     1
3     1
4     0
5     1
6     0
7     0
8     0
9     1
10    1
11    1
12    1
13    1
14    1
15    1
Name: salary_more_then_100k, dtype: int64
```

train_test_split

In [19]:
```
from sklearn.model_selection import train_test_split
X_train,X_test,y_train,y_test=train_test_split(inputs_n,target,test_size=0.2)
```

In [20]:
```
len(X_train)
```

Out[20]:
```
12
```

In [21]:
```
len(X_test)
```

Out[21]:
```
4
```

Now we are ready to train our classifier.

In [22]:
```
from sklearn import tree
model = tree.DecisionTreeClassifier()
```

In [24]:
```
model.fit(X_train,y_train)
```

Out[24]:
```
DecisionTreeClassifier()
```

In [25]:
```
model.score(X_test,y_test)
```

Out[25]:
```
0.5
```

This is because the number of data is too small.

Prediction

Is salary of Google, Sales executive, Masters degree > 100 k ?

company	label
abc pharma	0
facebook	1
google	2

job	label
Business manager	0
Computer programmer	1
Sales executive	2

degree	label
bachelors	0
masters	1

In [27]:
```
model.predict([[2,2,1]])
```

Out[27]:
```
array([0], dtype=int64)
```

It means this person's salary is not greater than 100K.

Is salary of facebook, business manager, bachelor degree > 100 k ?

In [28]:
```
model.predict([[1,0,0]])
```

Out[28]:
```
array([1], dtype=int64)
```

It means that this persron's salary > 100K.

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Date	Author
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