In [67]:	<pre>df = pd.read_csv('breast_ca')</pre>	ancer.csv')									
In [68]:	df.head(10)										
Out[68]:	Sample code number Clump Thi										
	10000251002945	5 5	1 4	1	1 5	2 7	1	3	1 2	1 2 1 2	
	2 1015425 3 1016277	3 6	1 8	1 8	1 1	2	2 4	3	1 7	1 2 1 2	
	4 1017023	4	1	1	3	2	1	3	1	1 2	
	5 10171226 1018099	8	10 1	10 1	8 1	7	10 10	9	7	1 4 1 2	
	7 1018561 8 1033078	2	1	2 1	1 1	2	1 1	3 1	1	1 2 5 2	
	9 1033078	4	2	1	1	2	1	2	1	1 2	
In [69]:	df.shape										
Out[69]:	(683, 11)										
In [70]:	# ANOTHER METHOD TO DOWNLOA										
	<pre>df1 = pd.read_csv('https://archive.ics.uci.edu/ml/machine-learning-databases/breast-cancer-wisconsin/breast-cancer-wisconsin.data',</pre>										
In [71]:	df1.head(10)										
Out[71]:	Sample code number Clump Thi	ckness Unifor	mity of Cell Size Uniformity of	f Cell Shape Marginal A	Adhesion Single Ep	ithelial Cell Size Bare	Nuclei Bland	Chromatin Normal	Nucleoli M	litoses Class	
	0 10000251 1002945	5 5	1 4	1	1 5	2 7	1	3	1 2	1 2 1 2	
	2 1015425	3	1	1	1	2	2	3	1	1 2	
	3 10162774 1017023	6 4	8 1	8 1	3	3 2	1	3	7	1 2 1 2	
	5 10171226 1018099	8	10 1	10 1	8	7 2	10 10	9	7	1 4 1 2	
	7 1018561	2	1	2	1	2	1	3	1	1 2	
	8 10330789 1033078	2	2	1	1	2	1	1 2	1	5 2 1 2	
In [72]:	des also										
	df1.shape										
Out[72]: In [73]:	(699, 11)										
[].	<pre>X = df.iloc[: , 1:-1].value y = df.iloc[: , -1].values</pre>	2 S									
In [74]:	<pre>from sklearn.model_selectic X train X test v train v te</pre>			=0 2 randa	= 0)						
T 51	<pre>X_train, X_test, y_train, y_test =train_test_split(X, y, test_size=0.2, random_state = 0)</pre>										
In [75]:	<pre>from sklearn.linear_model import LogisticRegression classifier = LogisticRegression(random_state = 0) classifier.fit(X_train, y_train)</pre>										
Out[75].	classifier.fit(X_train,y_train) LogisticRegression(random_state=0)										
Out[75]:	<pre>y_pred = classifier.predict(X_test)</pre>										
	<pre>print(np.concatenate((y_pred.reshape(len(y_pred),1),y_test.reshape(len(y_test),1)),1))</pre>										
	[[2 2] [2 2] [4 4]										
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	[2 2] [4 4] [2 2]]										
In [77]:	#applying confusion matrix	and checkin	g accuracy								
	<pre>from sklearn.metrics import cm = confusion_matrix(y_tes</pre>	confusion_ st,y_pred)	matrix,accuracy_score								
	print(cm) accuracy_score(y_test,y_pre										
	[[84 3] [3 47]]										
	0.9562043795620438										
In [78]:	<pre>#applying k-fold cross valuation from sklearn.model_selection</pre>										
	<pre>accuracies = cross_val_scor print('accuracy:{:.2f}%'.fo</pre>	re(estimator ormat(accura	<pre>= classifier, X=X, y=y cies.mean()*100))</pre>								
	print('standard deviation:{ accuracy:96.79%	[:.2f}%'.for	mat(accuracies.std()*10	00))							
	accuracy:96.79% standard deviation:2.50%										
In []:											

In [66]:

import pandas as pd
import numpy as np