	396 397	6.147.41	139 162	23 26	3 12	37 368	120 159	Medium Medium	40	11 18	No Ye Yes Ye	S						
	398 399	5.94 9.71	100 134	79 37	7	284 27	95 120	Bad Good	50 49	12 16	Yes Ye							
	400 ro	ws × 11 colu	mns															
<pre>In [3]: Out[3]:</pre>	coun		90000															
	mean std min 25%	2.82 0.00	96325 24115 90000 90000															
	50% 75% max	7.49 9.32 16.2	90000 20000 70000															
In [4]:		: Sales, di anging the = pd.get_c	categori	ical varia	bles into	dummi	ies.											
	#Cor	verting th	ne Target	t variable		es int	to Cat	egorical										
				bins=[0 labels=	,10, np.ir ['Low/Mid' _lowest =T r	', Hiç	gh'],											
Out[4]:	df1	Sales Comp	Price Inc	ome Advert	ising Popul	ation	Price	Age Educ	ation S	helveLoc_I	Bad Shelv	eLoc_Good Shelvel	_oc_Medium Urba	n_No Urba	n_Yes U	S_No US	_Yes Category	
	0	9.50 11.22	138 111	73 48	11 16	276 260	120 83	42 65	17 10		1 0	0 1	0	0	1	0	1 Low/Mid 1 High	
		10.06 7.40	113 117	35 100	10 4	269 466	80 97	59 55	12 14		0	0	1	0	1	0	1 High1 Low/Mid	
	4	4.15 		64	3	340	128	38	13 		1	0	0	0	1	1	0 Low/Mid	
	395 396	12.57 6.14	138 139	108 23	17 3	203 37	128 120	33 55	14 11		0	1 0	0	0	1	0	1 High1 Low/Mid	
	397 398	7.41 5.94	162 100	26 79	12 7	368 284	159 95	40 50	18 12		0	0	0	0	1	0	1 Low/Mid1 Low/Mid	
	399 400 ro	9.71 ws × 16 colu	134 Imns	37	0	27	120	49	16		0	1	0	0	1	0	1 Low/Mid	
In [5]:	# Rá	andom Fores	st Classi															
	from	n pandas i n n sklearn.n n sklearn.n n sklearn.e	nodel_sel nodel_sel	lection im lection im	port cross	s_val_												
In [6]:	arra	ay = df1.va	alues	тіїрог с ка	iluoiliFoi est	LOTASS	stitei											
	Y =	array[:,1: array[:,15	5]															
In [7]:	max_ kfo]	_trees = 10 _features = ld = KFold(el = Randon	= 4 (n_splits				ım tro	es may f	Feature	as=may fe	patures)							
	resu prin	ults = cros nt(results.	ss_val_sc	core(model				, max_1	- wear t	u^_1€								
	Let's try if we can increase thee cv score using ensemble techniques																	
In [8]:		gging n pandas in	nort rea	ad cev														
111 [0].	from from from	n sklearn.n n sklearn.n n sklearn.e	nodel_sel nodel_sel ensemble	lection im lection im import Ba	port cross ggingClass	s_val_ sifier	-											
	<pre>from sklearn.tree import DecisionTreeClassifier seed = 7</pre>																	
	<pre>cart = DecisionTreeClassifier() num_trees = 100 model1 = BaggingClassifier(base_estimator=cart, n_estimators=num_trees, random_state=seed)</pre>																	
	resu prin	ults1 = cro nt(results1	oss_val_s	score(mode				ocimacor c	J=114C		5 ca							
	86.25 R 0	osting																
In [9]:	# Ac	daBoost Cla																
	from	n pandas i n n sklearn.n n sklearn.n n sklearn.e	nodel_sel nodel_sel	lection im lection im	port cross	s_val_												
	mode	el2 = AdaBo ults2 = cro	oostClass	sifier(n_e	stimators	=num_t	rees,	random_s	state=s	seed)								
		nt(results2																
	Sta	cking	П															
In [10]:	fron	tacking Ens n pandas i n n sklearn.n	n <mark>port</mark> rea	ad_csv		1												
	from from from	n sklearn.n n sklearn.] n sklearn.t	nodel_sel Linear_mo cree impo	lection <mark>im</mark> odel impor ort Decisi	port cross t Logistic	s_val_ cRegre	ession											
		n sklearn.s n sklearn.e			tingClassi	ifier												
		ation -																
In [11]:	esti mode	reate the s imators = [el3 = Logis imators.app	[] sticRegre	ession(max)												
	mode esti	el4 = Decis imators.app el5 = SVC()	sionTree(oend((' <mark>ca</mark>	Classifier	()													
	esti mode esti	imators.app el6 = Baggi imators.app	oend(('s\ ingClassi oend(('ba	ifier(base agging', m	_estimator odel6))						andom_sta	ite=seed)						
	esti	el7 = AdaBo imators.app	pend(('bo	oosting',	model7))	=num_t	rees,	random_s	state=s	seed)								
	ense resu	reate the e emble = Vot ults3 = cro nt(results3	ingClass	sifier(est score(ense	imators)	√, cv=	kfold)										
	90.5			,														
		ation-																
In [12]:	esti mode	reate the simators = [e18 = Logis	[] sticRegre	ession(max)												
	mode esti	imators.app el9 = Decis imators.app el10 = Bagg	sionTree(oend((' <mark>ca</mark>	Classifier art', mode	() 19)))r=car	t, n	estimator	S=NIIm	_trees _r	andom st	ate=seed)						
	esti mode	e110 = Bagç imators.app e111 = AdaE imators.app	pend((' <mark>ba</mark> BoostClas	<mark>agging'</mark> , m ssifier(n_	odel10)) estimators						S [5554)						
	# cr	reate the e	ensemble ingClass	model sifier(est	imators)	/	-b€∵-)										
	prin	ults4 = cro nt(results4 99999999999	1.mean()*		шите, X, \	, CV=	-v101q)										
		ation -																
In [13]:	esti	reate the simators = [[]		v. · ·													
	esti mode	el12 = Logi imators.app el13 = Deci imators.app	oend((' <mark>l</mark> o isionTree	o <mark>gistic['],</mark> eClassifie	model12)) r()	v)												
	mode esti	el14 = AdaE imators.app	BoostClas Dend(('bo	ssifier(n_ posting',	estimators	s=num_	_trees	, random_	_state=	seed)								
	ense resu	reate the emble = Vot ults5 = cro	ingClass	sifier(est score(ense		۲, cv=	kfold)										
	90.5	nt(results5	eall()*	±00)														
	Iter	ation-	4															
In [14]:	esti mode	reate the s imators = [el15 = Deci] isionTree	eClassifie														
	mode	imators.app el16 = AdaE imators.app	BoostClas	ssifier <mark>(</mark> n_	estimators	s=num_	_trees	, random_	_state=	seed)								
	ense	reate the e emble = Vot ults6 = cro	ingClass	sifier(est		(, cv-	kfold)										
		nt(results6			, 1		J±U											
	Iter	ation-	5															
In [15]:	esti	reate the s imators = [[]		v ito	a)												
	esti mode	el17 = Logi imators.app el18 = AdaE imators.app	oend((' <mark>l</mark> o BoostClas	o <mark>gistic'</mark> , ssifier(n_	model17)) estimators		_trees	, random_	_state=	seed)								
	# C1	imators.app reate the e emble = Vot	ensemble	model														
	resu prin	ults6 = cro	oss_val_s	score(ense		√, cv=	kfold)										
	90.0 Sin	ice the	; CN e	core f	or iter	atic)n 3	was	the	mav	. SO V	ve can co	nsider it	to he	Our f	final i	model	
In []:	ااات	ioc uit	. UV S	OUIC I	טו ונכו	ull) I I	vvaS	แเน	παχ	, 30 V	ve can cu	i ioiuti Il	io ne	JUI	mal l	HOUEI	
-																		

import pandas as pd
import numpy as np
import warnings
warnings.filterwarnings("ignore")

In [2]: df=pd.read_csv("/Users/SAURABH/Saurabh patil/DATA SCIENCE/random Forest/Company_Data.csv")

Sales CompPrice Income Advertising Population Price ShelveLoc Age Education Urban US

340 128

203 128

Bad 42

Good 65

Medium 59

Medium 55

Bad 38

Good 33

17 Yes Yes

10 Yes Yes

13 Yes No

14 Yes Yes

Yes Yes

Yes Yes

In [1]:

Out[2]:

df

9.50

11.22

10.06

7.40

4.15

12.57