IN [I]:	impor impor impor impor from	rt random rt pandas as r rt numpy as n rt matplotlib rt seaborn as sklearn.mode	p .pyplot <b>as</b> sns l_selectio	n <b>import</b> tra						
In [2]:	from	sklearn.enses sklearn.neig	hbors <b>impo</b>	rt KNeighbon	rsClassifie		R)\\Breast	cancer data	CSV")	
0.1503	df							cancer_data		
Out[2]:	0	17.99	10.38	122.80	1001.0	mean_smoothness 0.11840	0			
	1 2	20.57 19.69	17.77 21.25	132.90 130.00	1326.0 1203.0	0.08474 0.10960	0			
	3	11.42 20.29	20.38 14.34	NaN NaN	386.1 1297.0	0.14250 0.10030	0			
	564 565	NaN NaN	22.39 28.25	142.00 131.20	NaN NaN	0.11100 0.09780	0			
	566 567	16.60 20.60	28.08 29.33	108.30 140.10	NaN NaN	0.08455 0.11780	0			
	568	7.76	24.54	47.92	NaN	0.05263	1			
	569 row	vs × 6 columns								
In [3]:		ck the null vsnull().sum()	alue in da	taset						
Out[3]:	mean_r mean_r mean_s diagno	smoothness	23 51 39 29 440 0							
In [24]:		l null value df.fillna(df.								
Out[24]:	0	nean_radius mea	n_texture m	ean_perimeter		mean_smoothness 0.11840	<b>diagnosis</b>			
	1	20.570000	17.77	132.900000	1326.00000	0.08474	0			
	3	19.690000 11.420000	21.25	130.000000 92.095604	386.10000	0.10960 0.14250	0			
	4	20.290000	14.34	92.095604	1297.00000 	0.10030	0			
	564 565	14.076482 14.076482	22.39 28.25	142.000000 131.200000	658.94463 658.94463	0.11100 0.09780	0			
	566	16.600000	28.08	108.300000	658.94463	0.08455	0			
	567 568	20.600000 7.760000	29.33 24.54	140.100000 47.920000	658.94463 658.94463	0.11780 0.05263	0			
	569 row	vs × 6 columns								
In [25]:		pairplot(df1,			-0.5					
Out[25]:		orn.axisgrid.	rairGrid a	c ux2d0be40	ayaU> 	a				
	25 snipe 20			0.0 O.0.		. Approximate		Re de.		
	mean radius		-		•					
	10		[			' <b>:</b>	/:			
	30 e 25									
	mean_texture 20	-								
	10									
	180 160 140 120			100 mg						
	120 100 80		-				a state of the sta			diagnosis 0 1
	60 40 2500									
	2000		•		-	a.r.	-			
	1500 1000	and the state of t								
	500	are to the same of								
	0.16 S 0.14		-			•			1	
	0.10 mean 0.08	- co (@D) (\$300g) a	910 03				- C(2)	s) (0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	1	
	0.06	10 20	30	10 20 30	50	100 150 200	0 1000	2000 3000	0.05 0.10 0.15	
- 50.63		mean_radius		mean_texture		mean_perimeter		n_area	mean_smoothness	
In [26]:		l.drop('diagn L['diagnosis'		=1)						
In [27]:		in_test_split ain,X_test,y_		st=train tes	st_split(X	,y,test_size=0.	.2,random s	tate <b>=</b> 51)		
In [28]:			, <u>y_</u> te		(A)		,S	- 01/		
Out[28]:	X_tra (455,	ain.shape 5)								
In [29]:		sklearn.line		<b>mport</b> Linear	rRegression	n				
	lr.fi	inearRegressi it(X_train,y_ core(X_test,y	train)							
Out[29]:	0.4986	6730954981748	5							
In [30]:	class	sifier.fit(X_	train,y_tr	ain)	cimators=35	50,criterion='ç	gini')			
Out[30]:	class	sifier.score(5	X_test,y_t							
Out[30]: In [31]:				ssifier(n e	stimators=3	310,criterion='	'entropy')			
	class	sifier1.fit(X	_train,y_t	rain)			/			
Out[31]:	0.8684	421052631579								
In [32]:	knn.f	<pre>KNeighborsCla Eit(X_train,y score(X test,)</pre>	_train)	neighbors=5)						
Out[32]:		7368421052632	_							
In [33]:	from	sklearn.svm	import SVC							
	svc.f	Fit(X_train,y score(X_test,								
Out[33]:	0.877	1929824561403								
In [34]:		oredict(X_tes		0. 0 1 1	1 1 1	1. 1 1 1 1	1. 0 1			
Out[34]:	array	1, 1, 1, 1, 1, 0, 0, 0,	1, 1, 1, 1, 1, 1,	1, 0, 1, 1, 1, 0, 1, 0,	1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1	1, 1, 1, 1, 1, 1, 0, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1,	1, 1, 0, 0, 1, 1,			
			1, 1, 1,	1, 1, 0, 0,		0, 1, 0, 0, 0, 0, 0, 0, 1, 1, 1, 1, 1,				
In [35]:	y_tes	st								
Out[35]:	169 137 188	1 1 1								
	443 345	1 1								
	538 123 224	1 1 1								
	115 425 Name:	1 1 diagnosis, L	ength: 114	, dtype: in	t64					
In [37]:	#this	s model gives	89% accur	acy						
In [ ]:										

In [1]: import random

In [ ]: