Out[3]: Click here to toggle on/off the raw code.

Question 5

Loading Abalone dataset and printing the normalized dataset

Applying PCA as a pre-processing step on abalone dataset

Out[4]:		PC1	PC2	PC3
	0	-1.756019	-0.390532	-0.329928
	1	-3.362734	-0.105153	0.252264
	2	-0.482338	0.252055	-0.443918
	3	-1.509041	0.207608	-0.000519
	4	-3.654006	-0.272819	0.275035
	4172	0.801361	0.385426	-0.064832
	4173	0.719312	-0.329146	-0.293062
	4174	2.167373	0.724010	0.402521
	4175	1.647501	-0.305166	-0.306030
	4176	4.894542	-0.705798	0.550942

4177 rows × 3 columns

Applying LDA as a pre-processing step on abalone dataset

Out[5]:	0	
---------	---	--

Rings			
15	-0.791003	-0.235208	0.359351
7	-2.355522	0.336978	0.214024
9	0.766719	-0.246564	1.129422
10	-0.611434	0.098075	0.230542
7	-2.674301	0.527509	0.102575
11	0.921330	-0.612381	-0.272399
10	0.425796	-0.894428	-0.034727
9	1.064523	-0.385654	-0.787231
10	0.840757	-1.513723	-0.864217
12	0.843580	0.352389	-2.262564

4177 rows × 3 columns

Loading wine dataset

Out+	$\Gamma \subset I$	
out	[O]	•

	fixed acidity	volatile acidity	citric acid	residual sugar	chlorides	free sulfur dioxide	total sulfur dioxide	density	рН	sulphates	alcoho
0	7.0	0.270	0.36	20.7	0.045	45.0	170.0	1.0010	3.00	0.45	8.
1	6.3	0.300	0.34	1.6	0.049	14.0	132.0	0.9940	3.30	0.49	9.
2	8.1	0.280	0.40	6.9	0.050	30.0	97.0	0.9951	3.26	0.44	10.
3	7.2	0.230	0.32	8.5	0.058	47.0	186.0	0.9956	3.19	0.40	9.
4	7.2	0.230	0.32	8.5	0.058	47.0	186.0	0.9956	3.19	0.40	9.
95	7.1	0.260	0.29	12.4	0.044	62.0	240.0	0.9969	3.04	0.42	9.
96	6.0	0.340	0.66	15.9	0.046	26.0	164.0	0.9979	3.14	0.50	8.
97	8.6	0.265	0.36	1.2	0.034	15.0	80.0	0.9913	2.95	0.36	11.
98	9.8	0.360	0.46	10.5	0.038	4.0	83.0	0.9956	2.89	0.30	10.
99	6.0	0.340	0.66	15.9	0.046	26.0	164.0	0.9979	3.14	0.50	8.
400											

100 rows × 13 columns

Applying PCA as a pre-processing step on wine dataset

Out[7]:		PC1	PC2
	0	-2.185179	3.529983
	1	-0.247707	-0.553177
	2	-0.380592	0.365447
	3	-1.735882	0.929351
	4	-1.735882	0.929351
	6492	2.699833	-0.854172
	6493	2.524458	-1.161039
	6494	2.775507	-0.761733
	6495	2.984356	-0.767021
	6496	1.852698	-0.516246
	C407 "	0	

6497 rows × 2 columns

Applying LDA as a pre-processing step on wine dataset

(6497, 2)

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ι,		10	

		=
quality		
6	0.752078	-1.466209
6	1.445150	0.392049
6	-0.123015	0.911451
6	0.288961	-0.721769
6	0.288961	-0.721769
5	0.512278	-0.224430
6	-0.514707	-0.597340
6	-0.231160	-0.831907
5	0.630811	0.158871
6	-0.668993	-2.296580

6497 rows × 2 columns

Gradient Boosting on Abalone dataset

Printing Test accuracy for raw abalone dataset

Plotting confusion matrix for abalone raw dataset

Classification Report:

	precision	recall	f1-score	support
1	0.00	0.00	0.00	0
3	0.00	0.00	0.00	3
4	0.40	0.31	0.35	13
5	0.42	0.34	0.38	32
6	0.28	0.25	0.26	48
7	0.27	0.26	0.27	84
8	0.26	0.33	0.29	99
9	0.25	0.29	0.27	142
10	0.27	0.32	0.29	139
11	0.23	0.25	0.24	93
12	0.18	0.12	0.14	51
13	0.11	0.13	0.12	31
14	0.20	0.04	0.06	26
15	0.00	0.00	0.00	21
16	0.00	0.00	0.00	13
17	0.00	0.00	0.00	8
18	0.00	0.00	0.00	12
19	0.00	0.00	0.00	7
20	0.00	0.00	0.00	4
21	0.00	0.00	0.00	3
22	0.33	0.33	0.33	3
23	1.00	0.25	0.40	4
24	0.00	0.00	0.00	0
accuracy			0.24	836
macro avg	0.18	0.14	0.15	836
weighted avg	0.24	0.24	0.24	836

Confusion Matrix:

[[0	0	0	(9 6	9 6) (9 6	9 (0	0	0	0	0	0	0	0	0	0	0	0	0	0]
[1	0	1	0	1	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0]
[0	2	4	5	1	1	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0]
[0	3	5	11	7	6	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0]
[0	0	0	6	12	17	8	5	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0]
[0	0	0	3	16	22	20	13	5	1	1	0	0	0	1	0	1	0	1	0	0	0	0]
[0	0	0	1	3	15	33	28	11	4	2	0	0	0	0	0	0	0	0	0	2	0	0]
[0	0	0	0	2	6	30	41	35	20	1	4	0	0	0	0	3	0	0	0	0	0	0]
[0	0	0	0	0	8	17	37	44	19	8	2	0	0	0	2	2	0	0	0	0	0	0]
[0	0	0	0	0	3	9	17	26	23	5	6	0	1	1	0	1	0	0	1	0	0	0]
[0	0	0	0	1	2	3	4	16	10	6	4	1	2	0	0	1	0	0	0	0	0	1]
[0	0	0	0	0	2	3	2	6	9	1	4	0	1	2	0	1	0	0	0	0	0	0]
[0	0	0	0	0	0	2	6	7	3	1	3	1	1	2	0	0	0	0	0	0	0	0]
[0	0	0	0	0	0	0	2	9	1	1	7	1	0	0	0	0	0	0	0	0	0	0]
[0	0	0	0	0	0	0	1	1	4	3	1	0	1	0	0	1	0	1	0	0	0	0]
[0	0	0	0	0	0	0	3	1	2	0	0	1	0	0	0	1	0	0	0	0	0	0]
[0	0	0	0	0	0	0	1	1	4	1	1	1	2	0	0	0	1	0	0	0	0	0]
[0	0	0	0	0	0	0	0	1	1	1	2	0	1	0	0	0	0	1	0	0	0	0]
[0	0	0	0	0	0	0	0	1	0	1	2	0	0	0	0	0	0	0	0	0	0	0]
[0	0	0	0	0	0	0	0	1	0	0	1	0	1	0	0	0	0	0	0	0	0	0]
Ĺ	0	0	0	0	0	0	0	1	0	0	1	0	0	0	0	0	0	0	0	0	1	0	0]
[0	0	0	0	0	0	0	0	0	1	1	0	0	0	1	0	0	0	0	0	0	1	0]
[0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0]]

Training Score: 1.0

Testing Score: 0.24282296650717702

The accuracy on the abalone raw dataset using Gradient Boosting classifier is less than Random Forests when using similar parameters, possibly due to the effect of outliers. It takes longer to train with Gradient Boosting than Random Forests.

Lack of strong linear relationships: Gradient boosting relies on creating ensembles of weak learners, which are typically decision trees, to model complex non-linear relationships between the input features and the target variable. However, if the input features do not exhibit strong linear relationships with the target variable, the decision trees may not be able to capture the complex non-linear relationships in the data.

Insufficient number of features: The Abalone dataset contains only 8 input features, which may not be sufficient to capture all the complexity in the data. If the input features do not provide enough information to accurately predict the target variable, the model may not perform well.

Overfitting: Gradient boosting can be prone to overfitting if the hyperparameters are not tuned properly. Overfitting occurs when the model learns to fit the training data too closely, which can lead to poor generalization performance on new, unseen data. This can happen if the model is too complex relative to the size of the training data, or if the learning rate is set too high, which causes the model to over-emphasize the contribution of individual trees.

Randomness in the data: The Abalone dataset contains some randomness due to the nature of the abalone shells and the way they grow. This can make it difficult for any model to accurately predict the age of an abalone based solely on physical measurements, which may contribute to the poor performance of gradient boosting on this dataset.

Changing the parameter: n_estimator=100, lr=0.1, and max depth 3 for raw abalone dataset

priting the confusion matrix after changing the parameters

Classification Report:

	precision	recall	f1-score	support
3	0.40	0.67	0.50	3
4	0.35	0.46	0.40	13
5	0.48	0.47	0.48	32
6	0.28	0.27	0.28	48
7	0.28	0.27	0.28	84
8	0.25	0.33	0.29	99
9	0.29	0.35	0.32	142
10	0.27	0.28	0.27	139
11	0.26	0.25	0.26	93
12	0.04	0.02	0.03	51
13	0.10	0.10	0.10	31
14	0.21	0.15	0.18	26
15	0.00	0.00	0.00	21
16	0.09	0.08	0.08	13
17	0.25	0.12	0.17	8
18	0.00	0.00	0.00	12
19	0.00	0.00	0.00	7
20	0.00	0.00	0.00	4
21	0.00	0.00	0.00	3
22	0.00	0.00	0.00	3
23	0.00	0.00	0.00	4
24	0.00	0.00	0.00	0
25	0.00	0.00	0.00	0
accuracy			0.25	836
macro avg	0.15	0.17	0.16	836
weighted avg	0.24	0.25	0.25	836

Confusion Matrix:

[[2	1	. 6	9 6	9 6	9 6	9 6	9 6	9 0	0	0	0	0	0	0	0	0	0	0	0	0	0	0]
[2	6	4	1	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0]
[1	8	15	5	3	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0]
[0	2	6	13	20	5	2	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0]
[0	0	5	19	23	24	9	1	0	0	0	0	0	1	0	1	0	1	0	0	0	0	0]
[0	0	1	2	15	33	35	9	0	1	0	0	1	1	0	0	0	0	0	1	0	0	0]
[0	0	0	3	10	29	49	35	9	3	1	2	0	0	0	0	1	0	0	0	0	0	0]
[0	0	0	2	4	23	29	39	21	8	3	1	4	2	0	1	2	0	0	0	0	0	0]
[0	0	0	0	5	6	19	27	23	4	5	2	0	0	0	2	0	0	0	0	0	0	0]
[0	0	0	0	1	6	6	13	12	1	4	4	1	1	0	0	0	1	0	0	0	1	0]
[0	0	0	1	1	2	3	7	7	3	3	0	2	2	0	0	0	0	0	0	0	0	0]
[0	0	0	0	0	2	9	2	5	0	2	4	0	1	0	0	1	0	0	0	0	0	0]
[0	0	0	0	0	0	3	6	1	1	4	3	0	1	0	0	0	1	0	0	1	0	0]
[0	0	0	0	0	0	0	2	1	2	2	1	1	1	1	0	0	2	0	0	0	0	0]
[0	0	0	0	0	1	1	1	2	1	0	1	0	0	1	0	0	0	0	0	0	0	0]
[0	0	0	0	0	0	2	1	3	2	2	0	1	0	0	0	1	0	0	0	0	0	0]
[0	0	0	0	0	0	0	1	1	1	2	0	0	0	0	0	0	1	1	0	0	0	0]
[0	0	0	0	0	0	0	0	1	0	2	0	0	0	1	0	0	0	0	0	0	0	0]
[0	0	0	0	0	0	0	0	1	0	0	0	1	0	1	0	0	0	0	0	0	0	0]
[0	0	0	0	0	0	1	1	0	0	0	0	1	0	0	0	0	0	0	0	0	0	0]
[0	0	0	0	0	0	0	0	0	1	0	1	0	1	0	0	0	0	0	0	0	0	1]
[0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0]
[0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0]]

Training Score: 0.6917090691409757 Testing Score: 0.25478468899521534

Upon using more optimum parameters for Gradient Boosting, the accuracy increases. This low accuracy may be due to the fact that the features are highly correlated.

Gradient Boosting on Wine dataset

Accuracy score: 0.5992

The accuracy of Gradient boosting on the wine - raw dataset is more than that of Random forests and this may be due to the fact that the dataset has outliers and is not balanced. When the dataset contains imbalanced classes, Random Forests may produce biased predictions towards the majority class, as each tree is built independently and can be influenced by the class imbalance, while Gradient Boosting Classifier can adjust the weights of the samples to balance the classes

Gradient Boosting on Abalone - PCA dataset

	precision	recall	f1-score	support
3	0.00	0.00	0.00	3
4	0.29	0.15	0.20	13
5	0.04	0.03	0.03	32
6	0.05	0.04	0.05	48
7	0.00	0.00	0.00	84
8	0.10	0.01	0.02	99
9	0.10	0.06	0.07	142
10	0.18	0.54	0.27	139
11	0.00	0.00	0.00	93
12	0.25	0.02	0.04	51
13	0.04	0.06	0.05	31
14	0.00	0.00	0.00	26
15	0.00	0.00	0.00	21
16	0.03	0.15	0.05	13
17	0.00	0.00	0.00	8
18	0.03	0.17	0.05	12
19	0.00	0.00	0.00	7
20	0.00	0.00	0.00	4
21	0.00	0.00	0.00	3
22	0.00	0.00	0.00	3
23	0.00	0.00	0.00	4
accuracy			0.11	836
macro avg	0.05	0.06	0.04	836
weighted avg	0.08	0.11	0.07	836

Confusion Matrix:

[[0	3	3 0	(9 0	(9 6	9 0	0	6	9 0	0	0	0	6	0	0	0	0	0	0]
[0	2	0	0	1	0	0	10	0	0	0	0	0	0	0	0	0	0	0	0	0]
[0	2	1	0	0	0	1	28	0	0	0	0	0	0	0	0	0	0	0	0	0]
[0	0	6	2	0	1	16	20	0	0	2	0	0	0	0	1	0	0	0	0	0]
[0	0	11	9	0	2	26	28	0	1	2	0	0	0	0	5	0	0	0	0	0]
[0	0	0	8	3	1	15	42	0	0	18	0	0	1	0	11	0	0	0	0	0]
[0	0	6	6	14	3	8	79	0	1	8	0	0	2	0	15	0	0	0	0	0]
[0	0	2	3	15	1	7	75	0	1	10	0	0	13	0	11	0	0	1	0	0]
[0	0	1	3	18	1	4	39	0	0	0	0	0	19	0	6	0	0	2	0	0]
[0	0	0	3	4	1	2	26	0	1	1	0	0	9	0	4	0	0	0	0	0]
[0	0	0	0	2	0	3	16	0	0	2	0	0	6	0	2	0	0	0	0	0]
[0	0	0	1	1	0	0	12	0	0	3	0	0	4	0	5	0	0	0	0	0]
[0	0	0	1	0	0	0	16	0	0	0	0	0	2	0	2	0	0	0	0	0]
[0	0	0	0	0	0	0	9	0	0	1	0	0	2	0	0	0	0	1	0	0]
[0	0	0	0	0	0	0	4	0	0	1	0	0	3	0	0	0	0	0	0	0]
[0	0	0	0	3	0	0	7	0	0	0	0	0	0	0	2	0	0	0	0	0]
[0	0	0	0	0	0	0	4	0	0	0	0	0	3	0	0	0	0	0	0	0]
[0	0	0	0	1	0	0	2	0	0	0	0	0	1	0	0	0	0	0	0	0]
[0	0	0	1	0	0	0	0	0	0	0	0	0	1	0	1	0	0	0	0	0]
[0	0	0	0	0	0	0	2	0	0	0	0	0	0	0	0	0	0	1	0	0]
[0	0	0	0	1	0	0	2	0	0	1	0	0	0	0	0	0	0	0	0	0]]

Training Score: 0.10356180784196349
Testing Score: 0.11483253588516747

The accuracy on PCA dataset upon using Gradient Boosting is lesser than Random forests. Overall it can be seen that PCA hurts the performance of a tree boosting classifier as data has been lost while reducing the number of dimensions.

Gradient Boosting on Wine - PCA dataset

Accuracy score: 0.5400

Classification Report:

	precision	recall	f1-score	support
3	0.00	0.00	0.00	2
4	0.32	0.22	0.26	46
5	0.57	0.56	0.57	420
6	0.56	0.63	0.59	579
7	0.50	0.38	0.43	221
8	0.35	0.25	0.29	32
accuracy			0.54	1300
macro avg	0.38	0.34	0.36	1300
weighted avg	0.54	0.54	0.54	1300

Confusion Matrix:

[[0	6) 2	2 0	0	0]
[1	10	10	21	2	2]
[2	10	237	158	10	3]
[3	7	141	362	61	5]
[1	3	24	103	85	5]
[0	1	3	8	12	8]]

Training Score: 0.8289397729459304

Testing Score: 0.54

The training score is 82% whereas the same classifer has a training score of approximately 70% on raw data without PCA reduction. So in this case, PCA helps in improving the accuracy but there is a considerable amount of overfitting.

Gradient Boosting on Abalone - LDA dataset

		- P			
		precision	recall	f1-score	support
	3	0.00	0.00	0.00	3
	4	0.00	0.00	0.00	13
	5	0.39	0.56	0.46	32
	6	0.30	0.29	0.29	48
	7	0.31	0.29	0.30	84
	8	0.20	0.31	0.24	99
	9	0.30	0.28	0.29	142
	10	0.18	0.19	0.18	139
	11	0.15	0.12	0.13	93
	12	0.38	0.06	0.10	51
	13	0.07	0.13	0.09	31
	14	0.00	0.00	0.00	26
	15	0.00	0.00	0.00	21
	16	0.25	0.23	0.24	13
	17	0.10	0.12	0.11	8
	18	0.00	0.00	0.00	12
	19	0.00	0.00	0.00	7
	20	0.00	0.00	0.00	4
	21	0.00	0.00	0.00	3
	22	0.00	0.00	0.00	3
	23	0.00	0.00	0.00	4
aco	curacy			0.21	836
macr	o avg	0.12	0.12	0.12	836
weighte	ed avg	0.21	0.21	0.20	836
_	_				

Confusion Matrix:

[[0	(9 3	3 (9 (9 6	9 6	9 6	0	6	0	0	0	0	0	0	0	0	0	0	0]
[0	0	13	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0]
[0	3	18	9	2	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0]
[0	3	4	14	14	8	1	1	0	0	0	0	1	0	0	0	0	1	0	0	1]
[0	0	5	12	24	32	9	0	2	0	0	0	0	0	0	0	0	0	0	0	0]
[0	0	1	4	17	31	26	14	1	0	1	1	2	0	0	0	1	0	0	0	0]
[0	0	1	3	12	30	40	34	17	0	4	0	1	0	0	0	0	0	0	0	0]
[0	0	0	2	6	27	26	27	20	2	13	8	2	2	3	0	0	0	0	0	1]
[0	0	1	0	2	8	21	31	11	1	10	0	4	2	1	0	0	1	0	0	0]
[0	0	0	1	0	5	5	13	8	3	8	3	3	0	1	1	0	0	0	0	0]
[0	0	0	1	1	2	2	13	3	0	4	1	2	2	0	0	0	0	0	0	0]
[0	0	0	1	0	4	1	7	3	1	4	0	1	2	1	0	0	0	0	0	1]
[0	0	0	0	0	3	1	6	3	0	6	1	0	0	0	0	1	0	0	0	0]
[0	0	0	0	0	0	0	2	2	0	0	5	0	3	0	0	1	0	0	0	0]
[0	0	0	0	0	0	1	1	1	1	0	1	0	0	1	1	0	1	0	0	0]
[0	0	0	0	0	2	1	2	2	0	2	0	0	1	1	0	0	0	0	0	1]
[0	0	0	0	0	1	0	2	1	0	1	1	0	0	0	1	0	0	0	0	0]
[0	0	0	0	0	1	0	0	0	0	0	2	0	0	1	0	0	0	0	0	0]
[0	0	0	0	0	1	0	0	1	0	1	0	0	0	0	0	0	0	0	0	0]
[0	0	0	0	0	3	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0]
[0	0	0	0	0	0	0	0	0	0	0	1	1	0	1	0	0	1	0	0	0]]

Training Score: 0.6240646513020054 Testing Score: 0.21052631578947367 Training score is high as when the dataset has a small number of samples, Gradient boosting can overfit and since most features in the abalone dataset is highly correlated, dimensionality reduction has a positive effect on efficient computation. But testing score is very low as there is considerable loss of data and Gradient boosting works better with more features. The mean accuracy using Random Forests is 0.27 whereas for Gradient boosting, it is lower. This is possible if there are too many outliers/high correlation in the dataset, which is true for this case.

Gradient Boosting on Wine - LDA dataset

Accuracy score: 0.5531

Classification Repor	t	:
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	precision	recall	f1-score	support	
3	0.00	0.00	0.00	2	
4	0.33	0.09	0.14	46	
5	0.60	0.66	0.63	420	
6	0.54	0.64	0.59	579	
7	0.49	0.29	0.37	221	
8	0.20	0.03	0.05	32	
accuracy			0.55	1300	
macro avg	0.36	0.29	0.30	1300	
weighted avg	0.54	0.55	0.53	1300	

Confusion Matrix:

[[0	6) 2	20	0	0]
[0	4	26	15	1	0]
[2	4	279	131	4	0]
[1	3	152	370	49	4]
[2	1	5	148	65	0]
[0	0	0	18	13	1]]

Training Score: 0.6492207042524534 Testing Score: 0.553076923076923

There is less overfitting in the training data after using LDA and Gradient boosting techniques. The test accuracy is also close but not very high. Compared to random forests, the accuracy is similar on Wine - LDA dataset.