

## Raport 4 - Jakub Ner

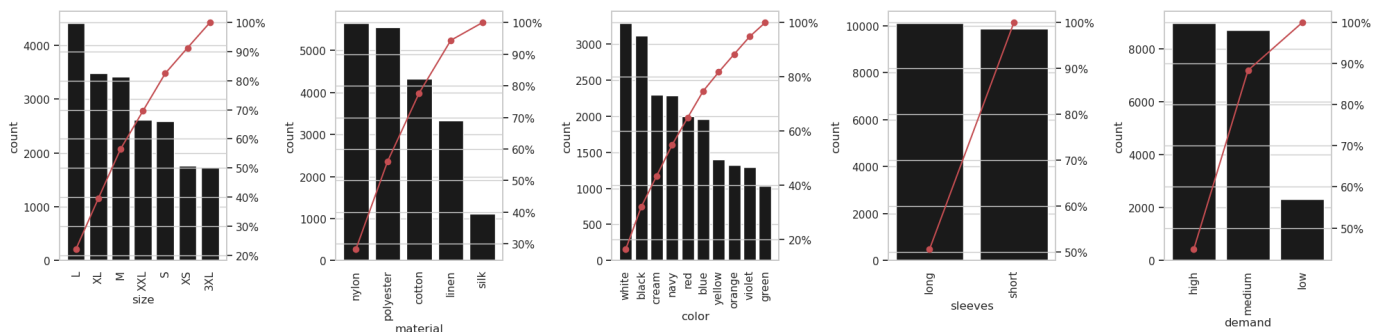
### 1. Data Exploration

I started for inspecting the data:

- There are 5 columns and 20 000 rows.
- There are no NaN values in the dataset.
- All values are categorical

	size	material	color	sleeves	demand
count	20000	20000	20000	20000	20000
unique	7	5	10	2	3
top	L	nylon	white	long	high
freq	4408	5652	3286	10117	8965

In order to examine data characteristics I plotted pareto charts for each feature



The key observation is that target values lack balance, with "low" values representing only 10% of the dataset.

### 2. Data Preprocessing

#### 2.1. One-hot encoding

I used one-hot encoding to convert categorical features to numerical. I dropped 1 level of each feature to avoid redundancy and multicollinearity.

One-hot (resp. one-cold) encoding creates co-linearity if all the features are used. Simply because the following relation always holds:  $\sum_i f_i = 1$ . So dropping one feature destroys the colinearity and it can have better results, since many models (esp. linear models) get confused with colinearities in the features. source: <https://datascience.stackexchange.com/questions/96526/two-questions-about-one-hot-encoding-drop-first-and-features-with-thousands-of>

#### 2.2. Feature standardization

I used StandardScaler to standardize the columns according to the formula  $z = (x - u) / s$  where  $u$  is the mean of the training samples and  $s$  is the standard deviation of the training samples. This operation assumes that the data is normally distributed within each feature and scales them such that the distribution is now centred around 0, with a standard deviation of 1.

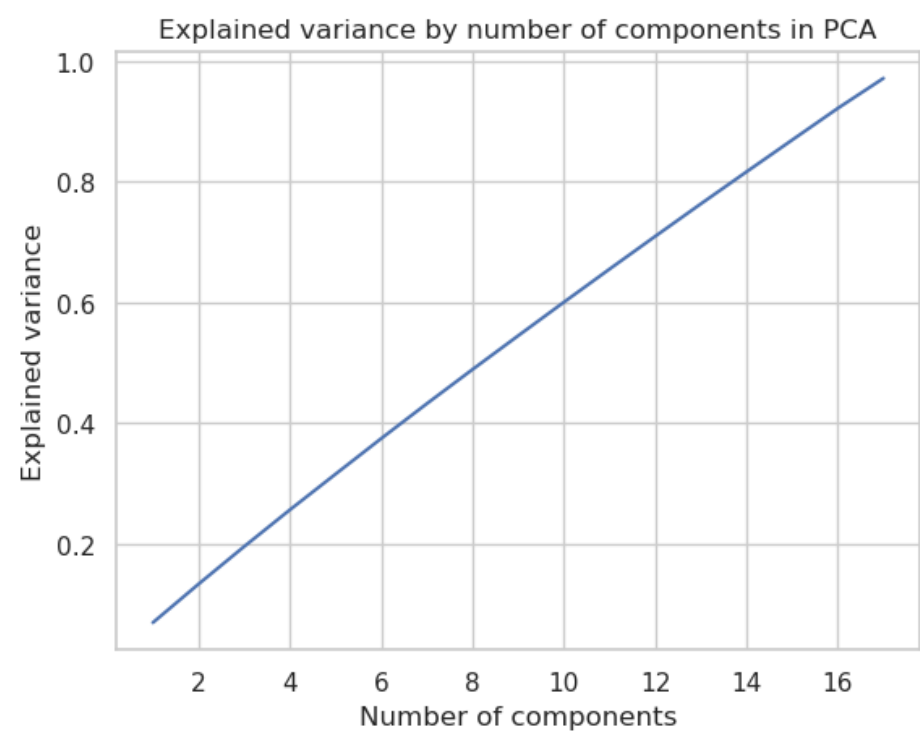
#### 2.3. Samples normalization

I scaled each sample by dividing its values by euclidean norm (l2) of the sample. This operation ensures that the samples are on the same scale.

*Note: One hot encoded categorical values does not contains diversified weights (the values are 0 or 1). Moreover, it refers to all features in the dataset. Therefore those are not important for the model itself, but the PCA requires all predictors to be on the same scale.*

#### 2.4. Dimensionality reduction - PCA

To reduce number of features I used Principle Component Analysis to explain 95% of variance. This lead to dimensionality reduction from 20 to 17 features.



3. Classification

3.1. Gaussian Naive Bayes

Gaussian Naive Bayes is a simple probabilistic classifier based on applying Bayes' theorem with naive independence assumptions between the features.

3.1.1. Impact of normalization and standarization

I compared the results of the model with and without normalization and standarization. For this benchmark I didn not utilize PCA. Standarization and normalization, when applied separetely, did not have a significant impact on the model performance. However, when applied together and in the correct order - first standarization, then normalization, the model accuracy improved by 11 percentage points. After standarization, all features have the same variance, and normalization further adjusts the scale without altering the standardized distribution significantly.

Without Normalization and Standarization

Accuracy: 0.50275

Confusion Matrix:

	Predicted: Low	Predicted: Medium	Predicted: High
Actual: Low	1803	0	0
Actual: Medium	259	183	15
Actual: High	1701	14	25

Classification Report:

	precision	recall	f1-score	support
high	0.479	1.0	0.648	1803.0
low	0.929	0.4	0.56	457.0
medium	0.625	0.014	0.028	1740.0
accuracy	0.503	0.503	0.503	0.503
macro avg	0.678	0.472	0.412	4000.0
weighted avg	0.594	0.503	0.368	4000.0

With Normalization

Accuracy: 0.5085

Confusion Matrix:

	Predicted: Low	Predicted: Medium	Predicted: High
Actual: Low	1803	0	0
Actual: Medium	234	214	9
Actual: High	1678	45	17

Classification Report:

	precision	recall	f1-score	support
high	0.485	1.0	0.653	1803.0
low	0.826	0.468	0.598	457.0
medium	0.654	0.01	0.019	1740.0
accuracy	0.508	0.508	0.508	0.508
macro avg	0.655	0.493	0.424	4000.0
weighted avg	0.598	0.508	0.371	4000.0

With Standarization

Accuracy: 0.49975

Confusion Matrix:

	Predicted: Low	Predicted: Medium	Predicted: High
Actual: Low	1803	0	0
Actual: Medium	271	171	15
Actual: High	1701	14	25

Classification Report:

	precision	recall	f1-score	support
high	0.478	1.0	0.646	1803.0
low	0.924	0.374	0.533	457.0
medium	0.625	0.014	0.028	1740.0
accuracy	0.5	0.5	0.5	0.5
macro avg	0.676	0.463	0.402	4000.0
weighted avg	0.593	0.5	0.364	4000.0

With normalization, then standarization

Accuracy: 0.50625

Confusion Matrix:

	Predicted: Low	Predicted: Medium	Predicted: High
Actual: Low	1803	0	0
Actual: Medium	243	205	9
Actual: High	1681	42	17

Classification Report:

	precision	recall	f1-score	support
high	0.484	1.0	0.652	1803.0
low	0.83	0.449	0.582	457.0
medium	0.654	0.01	0.019	1740.0
accuracy	0.506	0.506	0.506	0.506
macro avg	0.656	0.486	0.418	4000.0
weighted avg	0.597	0.506	0.369	4000.0

With Standarization, then normalization

Accuracy: 0.61875

Confusion Matrix:

	Predicted: Low	Predicted: Medium	Predicted: High
Actual: Low	1612	104	87
Actual: Medium	64	379	14
Actual: High	991	265	484

Classification Report:

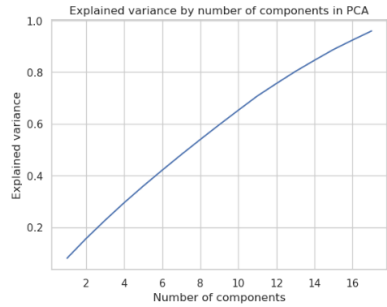
	precision	recall	f1-score	support
high	0.604	0.894	0.721	1803.0
low	0.507	0.829	0.629	457.0
medium	0.827	0.278	0.416	1740.0
accuracy	0.619	0.619	0.619	0.619
macro avg	0.646	0.667	0.589	4000.0
weighted avg	0.69	0.619	0.578	4000.0

3.1.2 Impact of PCA

Normalization and standarization make the result worse, because those cause that less features are dropped during the PCA. It impacts the model performance.

Standardized and normalized with PCA

Number of features before: 20 after: 17



Accuracy: 0.69725

Confusion Matrix:

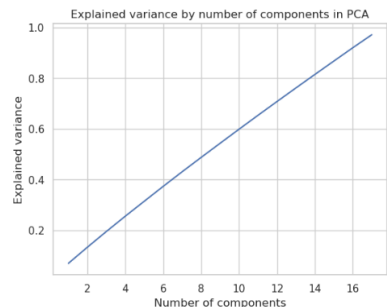
	Predicted: Low	Predicted: Medium	Predicted: High
Actual: Low	1535	88	180
Actual: Medium	67	268	122
Actual: High	563	191	986

Classification Report:

	precision	recall	f1-score	support
high	0.709	0.851	0.774	1803.0
low	0.49	0.586	0.534	457.0
medium	0.766	0.567	0.651	1740.0
accuracy	0.697	0.697	0.697	0.697
macro avg	0.655	0.668	0.653	4000.0
weighted avg	0.709	0.697	0.693	4000.0

Normalized, then standardized with PCA

Number of features before: 20 after: 17



Accuracy: 0.65675

Confusion Matrix:

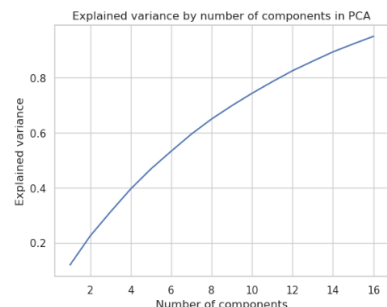
	Predicted: Low	Predicted: Medium	Predicted: High
Actual: Low	1312	60	431
Actual: Medium	55	261	141
Actual: High	504	182	1054

Classification Report:

	precision	recall	f1-score	support
high	0.701	0.728	0.714	1803.0
low	0.519	0.571	0.544	457.0
medium	0.648	0.666	0.626	1740.0
accuracy	0.657	0.657	0.657	0.657
macro avg	0.623	0.635	0.628	4000.0
weighted avg	0.657	0.657	0.656	4000.0

with PCA without normalization and standarization

Number of features before: 20 after: 16



Accuracy: 0.72975

Confusion Matrix:

	Predicted: Low	Predicted: Medium	Predicted: High
Actual: Low	1455	153	195
Actual: Medium	15	320	122
Actual: High	486	190	1144

Classification Report:

	precision	recall	f1-score	support
high	0.776	0.807	0.791	1803.0
low	0.483	0.7	0.571	457.0
medium	0.783	0.657	0.715	1740.0
accuracy	0.73	0.73	0.73	0.73
macro avg	0.68	0.722	0.692	4000.0
weighted avg	0.745	0.73	0.733	4000.0

3.2. Decision Tree

3.2.1. Impact of normalization and standarization

Normalization and standarization does not impact the model performance in terms of accuracy .

Decision trees do not require feature scaling or normalization, as they are invariant to monotonic transformations. They can also easily handle missing values and outliers, making them suitable for raw and noisy data.

Nevertheless, it can decrease number of nodes in the tree by over a half and decrease 2 levels of depth.

without normalization and standarization

Number of nodes: 689  
Depth of tree: 20

Standarized

Number of nodes: 689  
Depth of tree: 20

Normalized

Number of nodes: 331  
Depth of tree: 20

Normalized, then standarized

Number of nodes: 331  
Depth of tree: 20

Standarized, then normalized

Number of nodes: 257  
Depth of tree: 18

without normalization and standarization

Accuracy: 0.972

Confusion Matrix:

	Predicted: Low	Predicted: Medium	Predicted: High
Actual: Low	1781	0	22
Actual: Medium	0	421	36
Actual: High	31	23	1686

Classification Report:

	precision	recall	f1-score	support
high	0.983	0.988	0.985	1803.0
low	0.948	0.921	0.935	457.0
medium	0.967	0.969	0.968	1740.0
accuracy	0.972	0.972	0.972	0.972
macro avg	0.966	0.959	0.963	4000.0
weighted avg	0.972	0.972	0.972	4000.0

Standarized

Accuracy: 0.972

Confusion Matrix:

	Predicted: Low	Predicted: Medium	Predicted: High
Actual: Low	1781	0	22
Actual: Medium	0	421	36
Actual: High	31	23	1686

Classification Report:

	precision	recall	f1-score	support
high	0.983	0.988	0.985	1803.0
low	0.948	0.921	0.935	457.0
medium	0.967	0.969	0.968	1740.0
accuracy	0.972	0.972	0.972	0.972
macro avg	0.966	0.959	0.963	4000.0
weighted avg	0.972	0.972	0.972	4000.0

Normalized

Accuracy: 0.972

Confusion Matrix:

	Predicted: Low	Predicted: Medium	Predicted: High
Actual: Low	1781	0	22
Actual: Medium	0	421	36
Actual: High	31	23	1686

Classification Report:

	precision	recall	f1-score	support
high	0.983	0.988	0.985	1803.0
low	0.948	0.921	0.935	457.0
medium	0.967	0.969	0.968	1740.0
accuracy	0.972	0.972	0.972	0.972
macro avg	0.966	0.959	0.963	4000.0
weighted avg	0.972	0.972	0.972	4000.0

Normalized, then standardized

Accuracy: 0.972

Confusion Matrix:

	Predicted: Low	Predicted: Medium	Predicted: High
Actual: Low	1781	0	22
Actual: Medium	0	421	36
Actual: High	31	23	1686

Classification Report:

	precision	recall	f1-score	support
high	0.983	0.988	0.985	1803.0
low	0.948	0.921	0.935	457.0
medium	0.967	0.969	0.968	1740.0
accuracy	0.972	0.972	0.972	
macro avg	0.966	0.959	0.963	4000.0
weighted avg	0.972	0.972	0.972	4000.0

Standardized, then normalized

Accuracy: 0.972

Confusion Matrix:

	Predicted: Low	Predicted: Medium	Predicted: High
Actual: Low	1781	0	22
Actual: Medium	0	421	36
Actual: High	31	23	1686

Classification Report:

	precision	recall	f1-score	support
high	0.983	0.988	0.985	1803.0
low	0.948	0.921	0.935	457.0
medium	0.967	0.969	0.968	1740.0
accuracy	0.972	0.972	0.972	
macro avg	0.966	0.959	0.963	4000.0
weighted avg	0.972	0.972	0.972	4000.0

3.2.2 Impact of PCA

Similarly, PCA modifies only the structure of the model, what hypotetically for a bigger model may decrease chance of overfitting and improve inference time.

with PCA without normalization and standarization

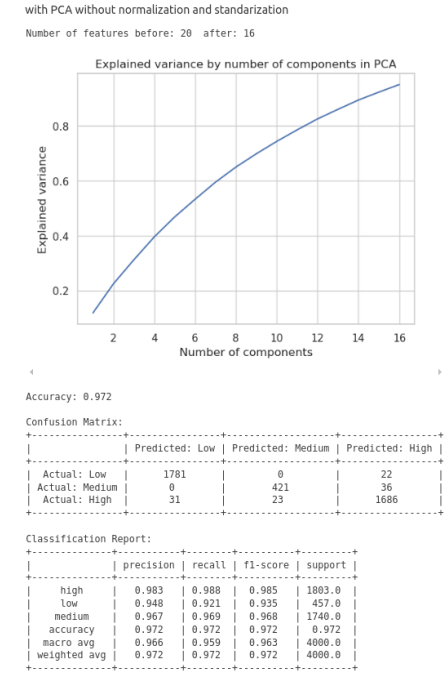
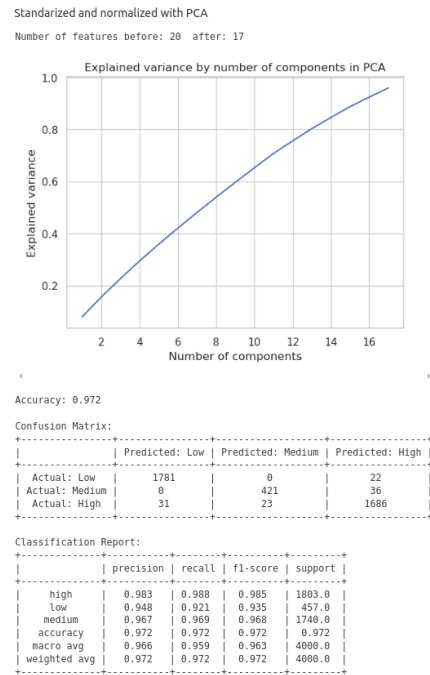
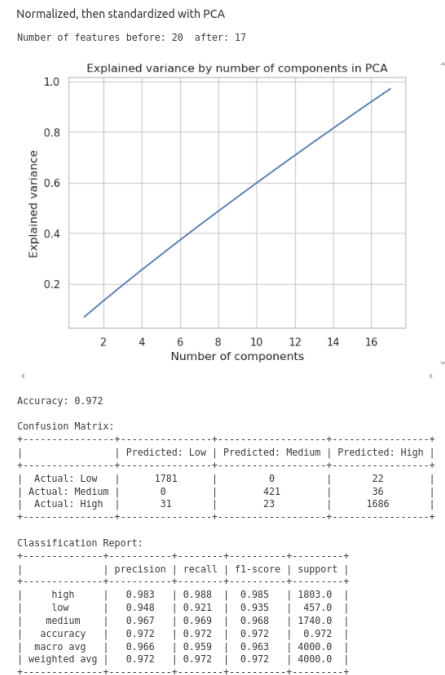
Number of features before: 20 after: 16  
Number of nodes: 257  
Depth of tree: 15

Standarized and normalized with PCA

Number of features before: 20 after: 17  
Number of nodes: 227  
Depth of tree: 17

Normalized, then standardized with PCA

Number of features before: 20 after: 17  
Number of nodes: 279  
Depth of tree: 14



3.3. SVM

3.3.1. Impact of normalization and standarization

without normalization and standarization

Accuracy: 0.97075

Confusion Matrix:

	Predicted: Low	Predicted: Medium	Predicted: High
Actual: Low	1772	0	31
Actual: Medium	0	419	38
Actual: High	27	21	1692

Classification Report:

	precision	recall	f1-score	support
high	0.985	0.983	0.984	1803.0
low	0.952	0.917	0.934	457.0
medium	0.961	0.972	0.967	1740.0
accuracy	0.971	0.971	0.971	0.971
macro avg	0.966	0.957	0.962	4000.0
weighted avg	0.971	0.971	0.971	4000.0

Standardized

Accuracy: 0.97

Confusion Matrix:

	Predicted: Low	Predicted: Medium	Predicted: High
Actual: Low	1773	0	30
Actual: Medium	0	415	42
Actual: High	29	19	1692

Classification Report:

	precision	recall	f1-score	support
high	0.984	0.983	0.984	1803.0
low	0.956	0.908	0.932	457.0
medium	0.959	0.972	0.966	1740.0
accuracy	0.97	0.97	0.97	0.97
macro avg	0.966	0.955	0.96	4000.0
weighted avg	0.97	0.97	0.97	4000.0

Normalized

Accuracy: 0.971

Confusion Matrix:

	Predicted: Low	Predicted: Medium	Predicted: High
Actual: Low	1764	0	39
Actual: Medium	0	423	34
Actual: High	22	21	1697

Classification Report:

	precision	recall	f1-score	support
high	0.988	0.978	0.983	1803.0
low	0.953	0.926	0.939	457.0
medium	0.959	0.975	0.967	1740.0
accuracy	0.971	0.971	0.971	0.971
macro avg	0.966	0.96	0.963	4000.0
weighted avg	0.971	0.971	0.971	4000.0

Normalized, then standardized

Accuracy: 0.971

Confusion Matrix:

	Predicted: Low	Predicted: Medium	Predicted: High
Actual: Low	1766	0	37
Actual: Medium	0	417	40
Actual: High	23	16	1701

Classification Report:

	precision	recall	f1-score	support
high	0.987	0.979	0.983	1803.0
low	0.963	0.912	0.937	457.0
medium	0.957	0.978	0.967	1740.0
accuracy	0.971	0.971	0.971	0.971
macro avg	0.969	0.957	0.962	4000.0
weighted avg	0.971	0.971	0.971	4000.0

Standardized, then normalized

Accuracy: 0.9725

Confusion Matrix:

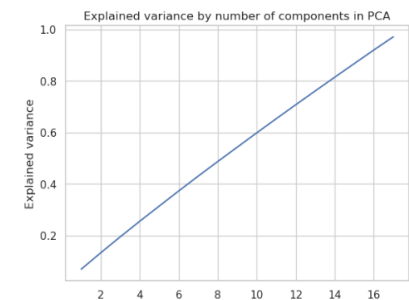
	Predicted: Low	Predicted: Medium	Predicted: High
Actual: Low	1774	0	29
Actual: Medium	0	424	33
Actual: High	29	19	1692

Classification Report:

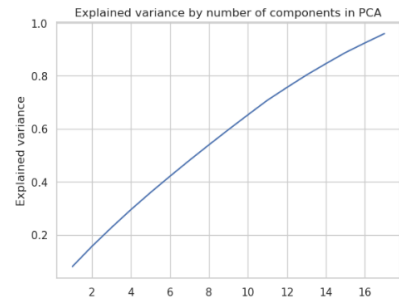
	precision	recall	f1-score	support
high	0.984	0.984	0.984	1803.0
low	0.957	0.928	0.942	457.0
medium	0.965	0.972	0.969	1740.0
accuracy	0.972	0.972	0.972	0.972
macro avg	0.969	0.961	0.965	4000.0
weighted avg	0.972	0.972	0.972	4000.0

3.3.2. Impact of PCA

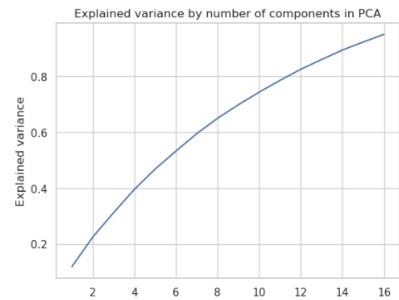
Normalized, then standardized with PCA  
Number of features before: 20 after: 17



Standardized and normalized with PCA  
Number of features before: 20 after: 17



with PCA without normalization and standarization  
Number of features before: 20 after: 16



Number of components				
Accuracy: 0.95325				
Confusion Matrix:				
	Predicted: Low	Predicted: Medium	Predicted: High	
Actual: Low	1769	0	34	
Actual: Medium	0	418	39	
Actual: High	98	16	1626	
Classification Report:				
	precision	recall	f1-score	support
high	0.948	0.981	0.964	1803.0
low	0.963	0.915	0.938	457.0
medium	0.957	0.934	0.946	1740.0
accuracy	0.953	0.953	0.953	0.953
macro avg	0.956	0.943	0.949	4000.0
weighted avg	0.953	0.953	0.953	4000.0

Number of components				
Accuracy: 0.95025				
Confusion Matrix:				
	Predicted: Low	Predicted: Medium	Predicted: High	
Actual: Low	1700	0	23	
Actual: Medium	0	403	54	
Actual: High	105	17	1618	
Classification Report:				
	precision	recall	f1-score	support
high	0.944	0.987	0.965	1803.0
low	0.96	0.882	0.919	457.0
medium	0.955	0.93	0.942	1740.0
accuracy	0.95	0.95	0.95	0.95
macro avg	0.953	0.933	0.942	4000.0
weighted avg	0.951	0.95	0.95	4000.0

Number of components				
Accuracy: 0.9435				
Confusion Matrix:				
	Predicted: Low	Predicted: Medium	Predicted: High	
Actual: Low	1706	0	97	
Actual: Medium	8	370	79	
Actual: High	20	22	1698	
Classification Report:				
	precision	recall	f1-score	support
high	0.984	0.946	0.965	1803.0
low	0.944	0.81	0.872	457.0
medium	0.906	0.976	0.94	1740.0
accuracy	0.944	0.944	0.944	0.944
macro avg	0.945	0.911	0.925	4000.0
weighted avg	0.945	0.944	0.943	4000.0

## Hyperparameters benchmarks

### Decision Tree

I tested the impact of the following hyperparameters on the normalized, then stadardized with PCA model:

1. max\_depth: 5 vs 10 vs 15 The best results are for max\_depth=15, but there is a change of overfitting.

max_depth=5				
Accuracy: 0.83325				
Confusion Matrix:				
	Predicted: Low	Predicted: Medium	Predicted: High	
Actual: Low	1597	43	163	
Actual: Medium	90	284	83	
Actual: High	177	111	1452	
Classification Report:				
	precision	recall	f1-score	support
high	0.857	0.886	0.871	1803.0
low	0.648	0.621	0.635	457.0
medium	0.855	0.834	0.845	1740.0
accuracy	0.833	0.833	0.833	0.833
macro avg	0.787	0.781	0.783	4000.0
weighted avg	0.832	0.833	0.833	4000.0
Number of nodes: 61				
Depth of tree: 5				

max_depth=10				
Accuracy: 0.97075				
Confusion Matrix:				
	Predicted: Low	Predicted: Medium	Predicted: High	
Actual: Low	1781	0	22	
Actual: Medium	3	420	34	
Actual: High	34	24	1682	
Classification Report:				
	precision	recall	f1-score	support
high	0.98	0.988	0.984	1803.0
low	0.946	0.919	0.932	457.0
medium	0.968	0.967	0.967	1740.0
accuracy	0.971	0.971	0.971	0.971
macro avg	0.964	0.958	0.961	4000.0
weighted avg	0.971	0.971	0.971	4000.0
Number of nodes: 255				
Depth of tree: 10				

max_depth=15				
Accuracy: 0.972				
Confusion Matrix:				
	Predicted: Low	Predicted: Medium	Predicted: High	
Actual: Low	1781	0	22	
Actual: Medium	0	421	36	
Actual: High	31	23	1686	
Classification Report:				
	precision	recall	f1-score	support
high	0.983	0.988	0.985	1803.0
low	0.948	0.921	0.935	457.0
medium	0.967	0.969	0.968	1740.0
accuracy	0.972	0.972	0.972	0.972
macro avg	0.966	0.959	0.963	4000.0
weighted avg	0.972	0.972	0.972	4000.0
Number of nodes: 279				
Depth of tree: 14				

2. criterion: entropy vs gini vs log\_loss Those hyperparameters affects number of nodes and depth of the tree. The best results are for gini criterion, because

criterion=entropy				
Accuracy: 0.972				
Confusion Matrix:				
	Predicted: Low	Predicted: Medium	Predicted: High	
Actual: Low	1781	0	22	
Actual: Medium	0	421	36	
Actual: High	31	23	1686	
Classification Report:				
	precision	recall	f1-score	support
high	0.983	0.988	0.985	1803.0
low	0.948	0.921	0.935	457.0
medium	0.967	0.969	0.968	1740.0
accuracy	0.972	0.972	0.972	0.972
macro avg	0.966	0.959	0.963	4000.0
weighted avg	0.972	0.972	0.972	4000.0
Number of nodes: 273				
Depth of tree: 16				

criterion=gini				
Accuracy: 0.972				
Confusion Matrix:				
	Predicted: Low	Predicted: Medium	Predicted: High	
Actual: Low	1781	0	22	
Actual: Medium	0	421	36	
Actual: High	31	23	1686	
Classification Report:				
	precision	recall	f1-score	support
high	0.983	0.988	0.985	1803.0
low	0.948	0.921	0.935	457.0
medium	0.967	0.969	0.968	1740.0
accuracy	0.972	0.972	0.972	0.972
macro avg	0.966	0.959	0.963	4000.0
weighted avg	0.972	0.972	0.972	4000.0
Number of nodes: 279				
Depth of tree: 14				

criterion=log_loss				
Accuracy: 0.972				
Confusion Matrix:				
	Predicted: Low	Predicted: Medium	Predicted: High	
Actual: Low	1781	0	22	
Actual: Medium	0	421	36	
Actual: High	31	23	1686	
Classification Report:				
	precision	recall	f1-score	support
high	0.983	0.988	0.985	1803.0
low	0.948	0.921	0.935	457.0
medium	0.967	0.969	0.968	1740.0
accuracy	0.972	0.972	0.972	0.972
macro avg	0.966	0.959	0.963	4000.0
weighted avg	0.972	0.972	0.972	4000.0
Number of nodes: 273				
Depth of tree: 16				

3. min\_samples\_split: 1 vs 3 vs 5 There is no significant difference between the results. It is worth noting that depth decreases with the increase of min\_samples\_split, because requiring more samples to split a node results in fewer splits overall, leading to a simpler and shallower tree structure

min_samples_leaf=1				
Accuracy: 0.972				
Confusion Matrix:				
	Predicted: Low	Predicted: Medium	Predicted: High	
Actual: Low	1781	0	22	
Actual: Medium	0	421	36	
Actual: High	31	23	1686	
Classification Report:				
	precision	recall	f1-score	support
high	0.983	0.988	0.985	1803.0
low	0.948	0.921	0.935	457.0
medium	0.967	0.969	0.968	1740.0
accuracy	0.972	0.972	0.972	0.972
macro avg	0.966	0.959	0.963	4000.0
weighted avg	0.972	0.972	0.972	4000.0
Number of nodes: 279				
Depth of tree: 14				

min_samples_leaf=3				
Accuracy: 0.97175				
Confusion Matrix:				
	Predicted: Low	Predicted: Medium	Predicted: High	
Actual: Low	1781	0	22	
Actual: Medium	0	421	36	
Actual: High	31	24	1685	
Classification Report:				
	precision	recall	f1-score	support
high	0.983	0.988	0.985	1803.0
low	0.946	0.921	0.933	457.0
medium	0.967	0.968	0.968	1740.0
accuracy	0.972	0.972	0.972	0.972
macro avg	0.965	0.959	0.962	4000.0
weighted avg	0.972	0.972	0.972	4000.0
Number of nodes: 273				
Depth of tree: 13				

min_samples_leaf=5				
Accuracy: 0.97175				
Confusion Matrix:				
	Predicted: Low	Predicted: Medium	Predicted: High	
Actual: Low	1781	0	22	
Actual: Medium	0	421	36	
Actual: High	31	24	1685	
Classification Report:				
	precision	recall	f1-score	support
high	0.983	0.988	0.985	1803.0
low	0.946	0.921	0.933	457.0
medium	0.967	0.968	0.968	1740.0
accuracy	0.972	0.972	0.972	0.972
macro avg	0.965	0.959	0.962	4000.0
weighted avg	0.972	0.972	0.972	4000.0
Number of nodes: 271				
Depth of tree: 12				

## References

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