## Beans

## December 16, 2022

```
In [65]: import pandas as pd
         from sklearn.model_selection import train_test_split
         import seaborn as sns
         from sklearn.ensemble import RandomForestClassifier
         import matplotlib.pyplot as plt
         df = pd.read_csv('/public/bmort/python/beans.csv')
         print(df.isnull().sum())
         print("There is missing data in the ShapeFactor3 column so lets fix this")
         imputed_value = df['ShapeFactor3'].median()
         df['ShapeFactor3'].fillna(imputed_value)
         df['ShapeFactor3'] = df['ShapeFactor3'].fillna(imputed_value)
         print("")
         print(df.describe())
         print("The maginitudes between the columns are much different, some are very small and
         print("The range between the columns varies a fair amount as well with some exception
         print("The range of area is large, the range of the convex area is also very large")
         print("It is clear to see there are outliers here because there is the max is so far
         sns.heatmap(df.corr(), cmap="YlGnBu")
         plt.show()
         print(df.corr())
         print("")
         train_x = df[['Area', 'Perimeter', 'AspectRatio', 'Eccentricity', 'roundness', 'Compactness'
         train_y = df['Class'].values
         print("the decision was taken to drop some features with correlations to area over .9
         print("the benefits of this decision should be a significant reduction of the computation
         print("I chose to use major and minor axis length")
         le = preprocessing.LabelEncoder()
         le.fit(df['Class'])
         le.transform(df['Class'])
```

```
train_x = df[['MajorAxisLength','MinorAxisLength','AspectRatio','Extent','Solidity','
         train_y = df['le_class'].values
         X_train, X_test, y_train, y_test = train_test_split(train_x, train_y, test_size=0.2)
         rf = RandomForestClassifier(n_estimators = 50)
         rf.fit(X_train, y_train);
         y_pred = rf.predict(X_test)
         print(rf.score(X_train, y_train))
         print(rf.score(X_test, y_test))
         from sklearn.metrics import confusion_matrix
         confusion_matrix(y_test, y_pred)
         from sklearn.metrics import ConfusionMatrixDisplay
         cm = confusion_matrix(y_test, y_pred)
         disp = ConfusionMatrixDisplay(confusion_matrix=cm)
         disp.plot()
         bean = pd.read_csv('/public/bmort/python/beans-unknown.csv')
         print(bean)
         test_bean = bean[['MajorAxisLength','MinorAxisLength','AspectRatio','Extent','Solidit
         bean_pred = rf.predict(test_bean)
         print("This is the bean prediction of the sample data")
         print(bean_pred)
                   0
Area
Perimeter
                   0
MajorAxisLength
                   0
MinorAxisLength
AspectRatio
Eccentricity
                   0
{\tt ConvexArea}
                   0
EquivDiameter
                   0
Extent
                   0
Solidity
                   0
roundness
                   0
Compactness
                   0
ShapeFactor1
                   0
ShapeFactor2
                   0
ShapeFactor3
                   1
ShapeFactor4
                   0
Class
                   0
dtype: int64
```

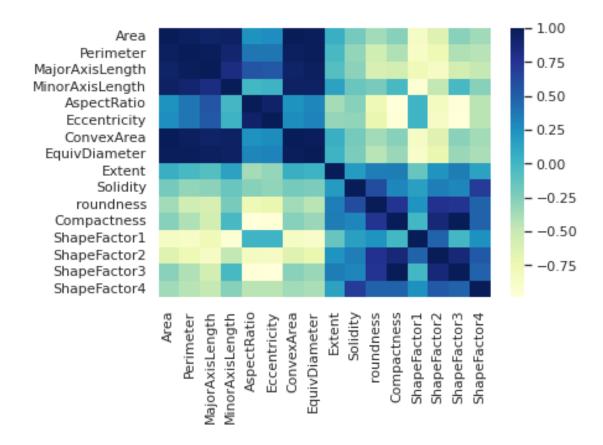
df['le\_class'] = le.transform(df['Class'])

There is missing data in the ShapeFactor3 column so lets fix this

count mean std min 25% 50% 75% max	Area 13533.000000 53057.388384 29401.235132 20420.000000 36269.000000 44581.000000 61422.000000 254616.000000	Perimeter 13533.000000 855.066479 214.781993 524.736000 703.180000 793.897000 977.266000 1985.370000	MajorAxisLeng 13533.0000 319.9229 85.8338 183.6011 253.0593 296.4413 376.3529 738.8603	13533. 981 202. 397 45. 165 122. 398 175. 382 192. 986 217.	~				
	${\tt AspectRatio}$	Eccentricity	ConvexArea	EquivDiamete	r Extent \				
count	13533.000000	13533.000000	13533.000000	13533.00000	0 13533.000000				
mean	1.581111	0.750310	53777.120890	253.05253	0 0.749827				
std	0.245328	0.091890	29853.152141	59.32488	6 0.048939				
min	1.024868	0.218951	20684.000000	161.24376	4 0.555315				
25%	1.430641	0.715134	36669.000000	214.89328	8 0.718753				
50%	1.549898	0.764011	45123.000000	238.24838	3 0.759903				
75%	1.704026	0.809699	62388.000000	279.65142	5 0.786847				
max	2.430306	0.911423	263261.000000	569.37435	8 0.866195				
	G 3 . 1	,	<b>a</b> .	G1	G1				
	Solidity	roundness	Compactness	ShapeFactor1	ShapeFactor2 \				
count	13533.000000	13533.000000	13533.000000	13533.000000	13533.000000				
mean	0.987150	0.873653	0.800347	0.006561	0.001719				
std	0.004651	0.059410	0.061485	0.001130	0.000596				
min	0.919246	0.489618	0.640577	0.002778	0.000564				
25%	0.985675	0.833360	0.763181	0.005890	0.001158				
50%	0.988287	0.883447	0.801505	0.006643	0.001700				
75%	0.990018	0.917039	0.834520	0.007271	0.002173				
max	0.994677	0.990685	0.987303	0.010451	0.003665				
	ShapeFactor3	ShapeFactor4							
count	13533.000000	13533.000000							
mean	0.644336	0.995077							
std	0.098687	0.004348							
min	0.410339	0.947687							
25%	0.582445	0.993717							
50%	0.642410	0.996393							
75%	0.696423	0.997891							
max	0.974767	0.999733							
The ma	The magnitudes between the columns are much different, some are very small and								

The maginitudes between the columns are much different, some are very small and some are large. The range between the columns varies a fair amount as well with some exceptions where they are. The range of area is large, the range of the convex area is also very large.

It is clear to see there are outliers here because there is the max is so far off of the mean.



	Area	Perimeter	${ t MajorAxisLength}$	MinorAxisLength	\
Area	1.000000	0.966904	0.932615	0.952038	
Perimeter	0.966904	1.000000	0.977558	0.914326	
MajorAxisLength	0.932615	0.977558	1.000000	0.828341	
MinorAxisLength	0.952038	0.914326	0.828341	1.000000	
AspectRatio	0.243660	0.386073	0.550062	-0.005404	
Eccentricity	0.268623	0.391125	0.541075	0.022423	
ConvexArea	0.999940	0.967868	0.933384	0.951777	
EquivDiameter	0.984997	0.991452	0.962267	0.949208	
Extent	0.054675	-0.020630	-0.077350	0.146002	
Solidity	-0.197216	-0.304551	-0.284758	-0.156688	
roundness	-0.358979	-0.548265	-0.595651	-0.213982	
Compactness	-0.269787	-0.407432	-0.567913	-0.018598	
ShapeFactor1	-0.848382	-0.865748	-0.775824	-0.947191	
ShapeFactor2	-0.641205	-0.768603	-0.859415	-0.475313	
ShapeFactor3	-0.273756	-0.408907	-0.567630	-0.022736	
ShapeFactor4	-0.357928	-0.431119	-0.484385	-0.266295	
	AspectRat	io Eccentr	icity ConvexArea	${ t Equiv Diameter}$	\
Area	0.2436	60 0.2	68623 0.999940	0.984997	
Perimeter	0.3860	73 0.3	91125 0.967868	0.991452	

MajorAxisLength	0.550062	0.	.541075	0	. 933384	0.962267	
MinorAxisLength	-0.005404	0.	.022423	0 /	. 951777	0.949208	
AspectRatio	1.000000	0.	.924207	0 /	. 245229	0.305206	
Eccentricity	0.924207	1.	.000000	0	. 270393	0.319410	
ConvexArea	0.245229	0.	. 270393	1.	.000000	0.985254	
EquivDiameter	0.305206	0.	.319410	0	. 985254	1.000000	
Extent	-0.371479	-0.	.319910	0	.052892	0.028773	
Solidity	-0.269104	-0.	. 298372	-0	. 206784	-0.232230	
roundness	-0.764988	-0.	.720220	-0	. 363531	-0.437107	
Compactness	-0.987647	-0.	.970317	-0,	. 271641	-0.328977	
ShapeFactor1	0.020914	0.	.017238	-0	.848374	-0.893397	
ShapeFactor2	-0.837337	-0.	.859269	-0	. 642770	-0.714696	
ShapeFactor3	-0.978534	-0.	.981064	-0,	. 275634	-0.331603	
ShapeFactor4	-0.451580	-0.	.450671	-0,	.364211	-0.394600	
	Extent Sc	lidity	roundn	ess	${\tt Compactness}$	ShapeFactor1	\
Area	0.054675 -0.	197216	-0.358	979	-0.269787	-0.848382	
Perimeter	-0.020630 -0.	304551	-0.548	265	-0.407432	-0.865748	
${ t MajorAxisLength}$	-0.077350 -0.	284758	-0.595	651	-0.567913	-0.775824	
${\tt MinorAxisLength}$	0.146002 -0.	156688	-0.213	982	-0.018598	-0.947191	
AspectRatio	-0.371479 -0.	269104	-0.764	988	-0.987647	0.020914	
Eccentricity	-0.319910 -0.	298372	-0.720	220	-0.970317	0.017238	
ConvexArea	0.052892 -0.	206784	-0.363	531	-0.271641	-0.848374	
EquivDiameter	0.028773 -0.	232230	-0.437	107	-0.328977	-0.893397	
Extent	1.000000 0.	192236	0.344	658	0.355158	-0.141615	
Solidity	0.192236 1.	000000	0.609	621	0.304833	0.154229	
roundness	0.344658 0.	609621	1.000	000	0.766030	0.234064	
Compactness	0.355158 0.	304833	0.766		1.000000	-0.005994	
ShapeFactor1	-0.141615 0.	154229	0.234	064	-0.005994	1.000000	
ShapeFactor2	0.237762 0.	344337	0.781	478	0.868350	0.473225	
ShapeFactor3	0.348469 0.	308660	0.761	057	0.998685	-0.005055	
ShapeFactor4	0.148651 0.	700132	0.472	688	0.486344	0.251063	
	ShapeFactor2	-	eFactor3		apeFactor4		
Area	-0.641205		).273756		-0.357928		
Perimeter	-0.768603		0.408907		-0.431119		
MajorAxisLength	-0.859415		).567630		-0.484385		
MinorAxisLength	-0.475313		0.022736		-0.266295		
AspectRatio	-0.837337		0.978534		-0.451580		
Eccentricity	-0.859269		0.981064		-0.450671		
ConvexArea	-0.642770		0.275634		-0.364211		
EquivDiameter	-0.714696		331603		-0.394600		
Extent	0.237762		348469		0.148651		
Solidity	0.344337		0.308660		0.700132		
roundness	0.781478		).761057		0.472688		
Compactness	0.868350		0.998685		0.486344		
ShapeFactor1	0.473225		0.005055		0.251063		
ShapeFactor2	1.000000	) (	0.872323		0.531714		

ShapeFactor3	0.872323	1.000000	0.486049
ShapeFactor4	0.531714	0.486049	1.000000

the decision was taken to drop some features with correlations to area over .9, the benefits of this decision should be a significant reduction of the computational complexity. I chose to use major and minor axis length

## 0.9995381489007944

## 0.9194680458071666

	Area	Perimeter	${ t MajorAxisLength}$	${ t MinorAxisLength}$	AspectRatio	\
0	37500	728.191	275.840463	173.818266	1.586948	
1	37500	715.578	272.171813	175.668301	1.549351	
2	37511	718.350	267.039757	179.141937	1.490660	
3	37513	720.028	269.589608	177.510928	1.518721	
4	37514	725.847	269.881174	177.418223	1.521158	

	Eccentricity	${\tt ConvexArea}$	${ t Equiv Diameter}$	Extent	Solidity	roundness	\
0	0.776481	37944	218.509686	0.703406	0.988299	0.888690	
1	0.763818	37797	218.509686	0.786229	0.992142	0.920295	
2	0.741599	37868	218.541732	0.717365	0.990573	0.913474	
3	0.752626	37981	218.547558	0.780545	0.987678	0.909270	
4	0.753547	37920	218.550471	0.793309	0.989293	0.894773	

	Compactness	ShapeFactor1	ShapeFactor2	ShapeFactor3	ShapeFactor4
0	0.792160	0.007356	0.001787	0.627517	0.995836
1	0.802837	0.007258	0.001860	0.644548	0.998631
2	0.818387	0.007119	0.001970	0.669756	0.998379
3	0.810668	0.007187	0.001915	0.657182	0.998076
4	0.809803	0.007194	0.001908	0.655780	0.997545

This is the bean prediction of the sample data

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