

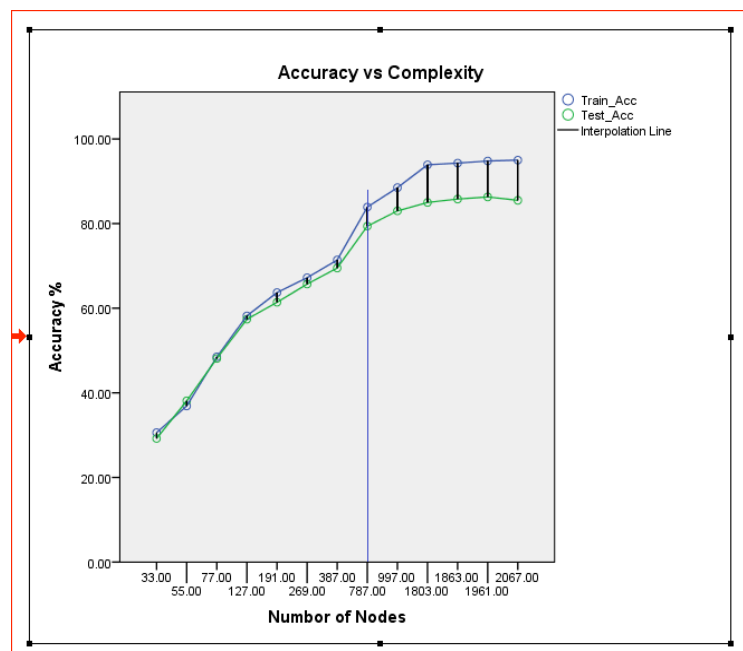
Question 1-a

Ans: The following shows the Accuracy of the Decision Tree model:

Variables:

- VAR00001 - capital letter (26 values from A to Z) -----Independent / Target Variable
- VAR00002 - x-box horizontal position of box (integer)
- VAR00003 - y-box vertical position of box (integer)
- VAR00004 - width width of box (integer)
- VAR00005 - high height of box (integer)
- VAR00006 - onpix total # on pixels (integer)
- VAR00007 - x-bar mean x of on pixels in box (integer)
- VAR00008 - y-bar mean y of on pixels in box (integer)
- VAR00009 - x2bar mean x variance (integer)
- VAR00010 - y2bar mean y variance (integer)
- VAR00011 - xybar mean x y correlation (integer)
- VAR00012 - x2ybr mean of $x * x * y$ (integer)
- VAR00013 - xy2br mean of $x * y * y$ (integer)
- VAR00014 - x-ege mean edge count left to right (integer)
- VAR00015 - xegvy correlation of x-ege with y (integer)
- VAR00016 - y-ege mean edge count bottom to top (integer)
- VAR00017 - yegvx correlation of y-ege with x (integer)

Model Summary		
Specifications	Growing Method	CRT
	Dependent Variable	VAR00001
	Independent Variables	VAR00002, VAR00003, VAR00004, VAR00005, VAR00006, VAR00007, VAR00008, VAR00009, VAR00010, VAR00011, VAR00012, VAR00013, VAR00014, VAR00015, VAR00016, VAR00017
	Validation	Split Sample
Results	Maximum Tree Depth	15
	Minimum Cases in Parent Node	10
	Minimum Cases in Child Node	5
	Independent Variables Included	VAR00012, VAR00008, VAR00011, VAR00007, VAR00010, VAR00014, VAR00013, VAR00009, VAR00015, VAR00005, VAR00003, VAR00002, VAR00016, VAR00006, VAR00017, VAR00004
	Number of Nodes	787
	Number of Terminal Nodes	394
	Depth	15



By observing the above Accuracy plot, the best model I would propose for this particular dataset including the following parameters, minimum of cases for parents is 10 and for child is 5, the stopping condition is with depth 15. In this model, the number of nodes is 797 with 394 terminal nodes. The sample split validation setting is 70% Training and 30% Testing data. The above result implies another feature that SPSS algorithm does push the tree to the maximum depth as preset.

Question 1-b: Report the misclassification matrix and interpret

		A	B	C	D	E	F
Training	A	501	1	0	7	0	0
	B	0	430	0	11	1	9
	C	1	3	478	0	11	6
	D	0	15	0	485	1	9
	E	0	2	6	1	427	3
	F	0	15	0	1	2	429
	G	2	5	9	8	13	6
	H	1	7	2	12	1	3
	I	0	18	2	1	0	8
	J	0	20	0	10	0	5
	K	0	4	6	6	4	0
	L	1	4	3	0	7	0
	M	5	0	0	6	0	1
	N	2	2	0	12	0	0
	O	0	12	0	18	0	0
	P	0	4	0	3	12	30
	Q	0	12	3	3	3	1
	R	5	23	2	10	4	1
	S	2	22	0	6	1	12
	T	0	7	3	3	7	0
	U	0	0	1	14	2	0
	V	0	1	0	2	0	3
	W	3	0	0	2	0	0
	X	0	18	2	1	7	5
	Y	0	5	4	0	0	4
	Z	1	10	1	5	10	10
	Overall Percent	3.80%	4.60%	3.80%	4.50%	3.70%	3.90%
Test	A	198	2	2	2	0	0
	B	0	177	0	3	1	7
	C	1	0	165	0	8	2
	D	1	12	0	173	0	1
	E	1	0	1	1	169	0
	F	0	5	0	0	1	194
	G	1	3	10	6	8	1
	H	1	5	2	8	1	2
	I	0	5	1	0	0	5
	J	0	7	0	1	0	2
	K	0	1	6	6	2	1
	L	1	1	3	0	4	0
	M	1	0	2	2	0	0
	N	1	0	0	3	0	1
	O	0	5	2	5	2	0
	P	0	0	0	2	4	14
	Q	2	4	0	1	3	1
	R	1	18	0	3	3	1
	S	2	18	0	8	2	1
	T	1	1	1	2	2	2
	U	0	0	3	6	0	0
	V	1	0	0	2	0	2
	W	2	0	0	0	0	1
	X	0	15	0	2	7	3
	Y	0	2	4	0	0	4
	Z	3	5	0	2	7	0
	Overall Percent	3.60%	4.70%	3.30%	3.90%	3.70%	4.00%

IS467
Homework#4
Date: 06-03-2017

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G	H	I	J	K	L	M	N
4	5	0	3	0	5	3	0
0	12	0	3	4	0	0	0
17	4	1	0	2	1	0	0
1	31	0	2	1	1	0	6
8	3	8	0	4	3	0	0
0	13	0	1	0	0	0	0
418	6	1	3	2	0	0	1
3	382	0	1	22	1	1	1
0	0	447	2	3	0	0	0
0	2	12	412	2	4	0	1
9	32	0	0	369	0	0	2
5	1	3	2	3	462	0	1
2	1	0	1	2	1	489	16
0	10	0	0	0	0	6	449
9	10	2	4	2	0	5	2
0	3	2	0	0	0	1	0
11	2	0	0	6	1	0	3
0	21	0	0	11	0	1	3
0	3	0	1	0	0	0	0
2	1	1	0	2	0	0	0
5	11	0	9	0	0	6	7
6	1	0	0	0	0	2	3
2	1	0	0	1	0	5	3
0	4	2	1	3	0	0	1
4	2	2	1	0	1	0	0
0	2	2	0	0	2	0	0
3.60%	4.10%	3.50%	3.20%	3.20%	3.50%	3.70%	3.60%
1	4	0	3	0	0	3	0
1	5	0	6	2	0	0	0
9	0	0	0	2	0	0	0
0	13	0	1	2	0	0	4
5	1	0	0	1	0	0	0
0	3	2	1	0	0	0	0
186	8	4	0	0	0	1	3
2	148	0	1	7	1	2	0
0	0	205	1	0	1	0	1
0	3	8	198	0	3	0	2
4	19	0	0	192	1	0	2
2	3	1	2	9	185	0	1
0	2	0	0	4	2	208	11
0	5	0	0	1	0	1	207
2	6	0	2	1	0	1	2
0	3	0	0	0	0	0	0
11	2	0	0	1	1	0	2
0	9	0	0	10	1	1	1
0	2	1	1	1	1	0	0
0	1	0	0	2	0	0	0
4	7	0	2	0	0	5	7
3	3	0	0	0	0	1	1
4	1	0	0	2	0	1	1
0	3	0	0	1	0	0	0
1	0	1	1	0	0	1	0
0	0	1	0	0	2	0	0
3.80%	4.10%	3.70%	3.60%	3.90%	3.20%	3.70%	4.00%

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Q	P	Q	R	S	T	U	V	W
0	0	1	8	4	1	2	7	0
1	2	2	6	8	1	0	31	0
6	1	2	3	3	0	0	0	0
7	4	0	5	2	1	1	4	0
2	0	4	0	9	2	2	4	0
1	18	0	0	10	8	0	8	4
5	2	2	2	2	0	0	14	2
11	3	0	18	3	1	3	19	2
1	8	0	3	14	3	0	0	0
0	1	1	0	14	1	0	0	0
0	0	0	11	0	2	2	9	0
4	0	5	3	5	1	0	1	0
2	0	1	1	1	0	2	10	6
10	5	1	2	0	0	5	32	1
430	4	9	1	1	0	2	14	4
1	500	0	0	0	0	0	32	1
26	6	408	8	7	1	0	3	1
2	6	0	396	5	1	0	24	0
3	6	2	2	409	0	0	4	0
0	0	0	4	8	500	0	1	3
10	1	1	0	0	0	466	15	4
0	7	0	0	0	3	1	503	9
0	0	2	0	0	0	1	22	475
3	0	0	4	0	0	0	4	0
2	1	1	0	1	18	1	38	0
3	1	1	1	9	3	0	0	0
3.80%	4.10%	3.20%	3.40%	3.70%	3.90%	3.50%	5.70%	3.70%
0	0	1	7	5	3	0	3	0
3	0	0	1	14	1	0	16	0
0	0	1	1	2	1	0	0	0
7	3	0	3	0	1	0	1	0
1	0	3	0	6	0	1	1	0
0	10	0	0	5	8	2	5	2
8	0	1	1	5	0	0	2	0
4	0	0	12	1	2	1	11	0
0	1	1	0	6	0	0	0	0
1	1	1	1	6	2	0	0	0
2	0	0	11	0	0	0	3	0
0	0	2	4	1	0	0	0	0
2	2	0	0	1	0	1	5	0
5	1	0	2	0	0	1	11	3
174	3	6	0	0	0	0	6	2
0	169	1	0	0	0	0	8	3
15	0	184	2	6	1	3	1	0
3	5	1	154	1	0	0	22	0
2	2	1	5	177	0	0	0	0
0	2	1	0	3	204	1	2	0
4	0	0	0	0	0	206	7	2
1	3	2	0	0	2	1	194	1
1	2	1	1	0	0	1	8	208
3	0	0	3	0	0	0	3	0
1	2	0	0	0	9	1	14	0
4	1	2	0	9	1	0	0	0
3.90%	3.40%	3.40%	3.40%	4.10%	3.80%	3.60%	5.30%	3.60%

X	Y	Z	Percent Correct
2	1	0	90.30%
5	0	0	81.70%
3	0	0	88.20%
1	1	1	83.80%
57	0	9	77.10%
15	4	0	81.10%
9	2	1	81.20%
13	3	2	74.20%
10	0	1	85.80%
14	2	0	82.20%
19	1	0	77.50%
20	0	1	86.80%
0	0	0	89.40%
0	3	0	83.10%
4	0	0	80.70%
2	3	1	84.00%
18	2	3	77.30%
3	0	1	76.30%
20	0	9	81.50%
10	9	2	88.80%
0	4	0	83.80%
0	3	0	92.50%
0	0	0	91.90%
480	0	1	89.60%
4	473	3	83.70%
4	2	438	86.70%
5.10%	3.70%	3.40%	83.90%
0	0	0	84.60%
3	0	0	73.80%
0	2	0	85.10%
2	1	0	76.90%
19	0	4	79.00%
5	3	0	78.90%
8	0	2	72.10%
7	1	0	67.60%
5	0	2	87.60%
7	1	2	80.50%
13	0	0	73.00%
10	0	0	80.80%
0	2	0	84.90%
0	1	0	85.20%
1	0	0	79.10%
1	3	0	81.30%
14	0	1	72.20%
5	0	0	64.40%
16	1	5	72.00%
3	3	2	87.60%
1	3	0	80.20%
0	3	0	88.20%
0	1	0	88.50%
210	1	0	83.70%
2	175	3	79.20%
5	2	185	80.80%
5.50%	3.30%	3.40%	79.40%

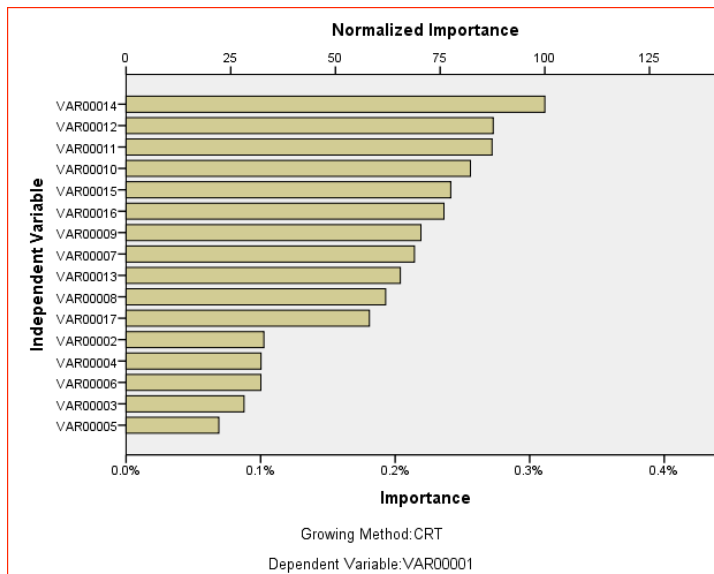
The reason I have picked this model because the accuracy of the Testing data (Blue Line) is 79.4% reaches the highest point of the curve with the least complexity (=787nodes). Also, the test set accuracy on the misclassification matrix are also acceptable range; the training and testing set accuracy ranges are very closed and the max percent accuracy of predicted value 1 reaches 88.5%. In addition, the standard deviation the 10-fold standard deviation (Shown below) is ONLY 0.005 with Estimate value 0.191 of the test set which is small enough to conclude that the model will not be overfitting. Also, this could imply that the Training and Testing data is split into an appropriate portion

(Training=70%/Testing=30%) for the classification model.

Risk		
Sample	Estimate	Std. Error
Training	.155	.003
Test	.191	.005
Growing Method: CRT		
Dependent Variable: VAR00001		

During the process of building the tree, I could observe that increasing the number of cases allowed in parent and child nodes is decreasing the complexity (number of nodes) of the tree. The mechanism of this effect because increasing the number of cases of nodes would let the node NOT splitting until it reaches the minimum of cases we set. So that the higher values we set on the parent and child nodes, the less complexity of tree we are supposed to get. Also, we could observe the above plot, the least of the complexity we have (e.g. Number of nodes =33), the less accuracy we could get.

Question 1-c: The most importance 3 attributes.



Independent Variable Importance		
Independent Variable	Importance	Normalized Importance
VAR00014	.311	100.0%
VAR00012	.273	87.7%
VAR00011	.272	87.4%
VAR00010	.256	82.3%
VAR00015	.241	77.5%
VAR00016	.236	75.9%
VAR00009	.219	70.4%
VAR00007	.214	68.9%
VAR00013	.204	65.5%
VAR00008	.193	62.0%
VAR00017	.181	58.1%
VAR00002	.102	32.9%
VAR00004	.100	32.2%
VAR00006	.100	32.2%
VAR00003	.088	28.2%
VAR00005	.069	22.1%

Growing Method: CRT
Dependent Variable: VAR00001

Based on the above, the most important three attributes are shown as the following in order:

- 1) VAR00014 - x-ege mean edge count left to right (integer) ---100% (Normalized Importance)
- 2) VAR00012 - x2ybr mean of x * x * y (integer) -----87.7% (Normalized Importance)
- 3) VAR00011 - xybar mean x y correlation (integer) -----87.4% (Normalized Importance)

Question 2-a:

Ans:

For this particular dataset, I have NOT transformed the Target variable to numeric number because they are single letters A-Z, there would be less chance to have errors in this case. On the other hand, the independent variables were transformed into smaller bin size(0-4=1,4-8=2,8-12=3,12-15=4) from the provided numerical integers. The reason to made this transformation is that 1.) reduce the algorithm running time 2.) To reduce the errors chances by the recorded numerical integers.

After running binned dataset, I tried to use the original dataset to run kNN again with the K (K=1,3,5,7). I found that the came up results are even more accurate. The partition ratio of the dataset is 70%Train / 30% Test data. The average accuracy of the original data result is ranged 94-96% and accuracy of the transformed dataset is ranged 78-80%. The detail matrix result is shown on part2-b. The reason might be the original data range are very standardized (1-15) for all the attributes. So that, the additional transformation / binning is not necessary for this case.

Question 2-b:

For the below shown matrix. The rows represent the provided labels count. The columns represent the predicted values count. The last column is the accuracy percentage of

Letter * Predicted Value for Letter Crosstabulation (K=1)																													
Count		Predicted Value for Letter																										Total	K=1
		A	B	C	D	E	F	G	H	I	J	K	L	M	N	O	P	Q	R	S	T	U	V	W	X	Y	Z		
Letter	A	786	0	1	0	0	0	0	0	0	1	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1	0	789	99.62%
	B	0	720	0	4	5	1	0	7	0	1	0	0	0	0	0	0	0	12	2	0	0	13	0	0	0	1	766	93.99%
	C	0	0	708	0	7	1	9	0	0	0	2	1	0	0	2	0	0	0	0	2	0	1	3	0	0	0	736	96.20%
	D	0	4	0	765	1	0	1	15	0	0	0	0	0	2	8	0	0	6	2	1	0	0	0	0	0	0	805	95.03%
	E	0	1	8	0	725	1	9	0	0	0	5	3	0	0	0	2	1	0	4	0	0	0	0	4	0	5	768	94.40%
	F	0	1	0	1	0	728	0	1	3	0	0	0	0	2	0	21	1	0	2	10	0	2	0	1	2	0	775	93.94%
	G	0	4	5	5	11	0	736	1	0	0	1	1	1	0	2	1	1	1	1	0	0	1	1	0	0	0	773	95.21%
	H	0	12	0	15	3	1	4	644	0	1	24	0	1	4	6	1	0	13	1	1	0	1	0	2	0	0	734	87.74%
	I	0	0	0	0	0	0	0	0	729	25	0	0	0	0	0	1	0	0	0	0	0	0	0	0	0	0	755	96.56%
	J	1	1	0	1	0	2	0	1	32	705	0	1	0	0	1	0	0	0	0	0	2	0	0	0	0	0	747	94.38%
	K	0	3	0	0	10	0	2	25	0	0	671	0	0	0	0	0	0	0	11	0	0	0	0	0	17	0	739	90.80%
	L	0	0	1	0	3	0	2	2	1	2	2	744	0	0	0	0	0	1	2	0	0	0	0	0	0	1	761	97.77%
	M	1	7	0	0	0	0	2	1	0	0	0	0	767	3	1	1	0	0	0	0	0	6	3	0	0	0	792	96.84%
	N	1	1	0	10	0	1	0	3	0	0	0	2	3	747	3	0	1	6	0	0	0	5	0	0	0	0	783	95.40%
	O	0	0	2	10	0	0	1	1	0	0	0	0	0	2	725	0	9	0	0	0	2	0	1	0	0	0	753	96.28%
	P	0	1	1	4	1	32	0	1	0	0	0	1	0	1	0	754	3	2	0	0	0	0	0	0	2	0	803	93.90%
	Q	0	1	0	2	1	0	1	0	0	0	0	0	0	0	18	3	752	2	1	0	0	0	0	0	1	1	783	96.04%
	R	0	19	0	2	0	2	0	10	0	0	15	2	0	6	0	0	1	701	0	0	0	0	0	0	0	0	758	92.48%
	S	0	5	0	1	6	0	1	2	0	1	0	1	0	0	0	0	1	0	725	1	1	0	0	0	0	3	748	96.93%
	T	0	3	2	1	0	3	0	1	1	1	0	0	0	0	0	0	0	2	0	769	0	0	0	0	12	1	796	96.61%
	U	2	0	1	1	0	0	0	7	0	0	3	0	2	0	0	0	0	0	0	0	796	1	0	0	0	0	813	97.91%
	V	0	12	1	0	0	0	1	0	0	0	0	0	2	2	1	2	0	0	0	0	0	740	1	0	2	0	764	96.86%
	W	0	0	0	0	0	0	1	0	0	0	0	0	3	1	3	0	0	1	0	0	2	1	740	0	0	0	752	98.40%
	X	0	1	0	3	4	0	0	0	1	0	10	0	0	0	2	0	1	1	1	1	0	0	0	760	0	2	787	96.57%
	Y	3	1	0	0	0	0	0	0	0	1	0	0	0	0	0	0	2	0	1	8	0	3	1	1	765	0	786	97.33%
	Z	0	0	0	0	3	1	0	0	0	0	0	0	0	0	0	0	7	0	2	1	0	0	0	0	0	720	734	98.09%
Total		794	797	730	825	780	773	770	722	767	738	733	756	779	770	772	786	781	760	742	794	803	774	750	785	785	734	20000	95.59%

Letter * Predicted Value for Letter Crosstabulation(K=3)																													
Count																													
		Predicted Value for Letter																										Total	
		A	B	C	D	E	F	G	H	I	J	K	L	M	N	O	P	Q	R	S	T	U	V	W	X	Y	Z		
Letter	A	779	0	2	0	0	0	0	0	0	0	0	1	2	1	0	0	0	0	0	1	1	0	0	0	2	0	789	98.73%
	B	0	720	0	9	6	2	0	2	0	0	1	0	2	1	0	0	0	10	1	0	1	8	0	1	1	1	766	93.99%
	C	1	0	695	0	7	0	11	0	0	0	0	1	0	0	8	0	2	0	0	1	7	0	3	0	0	0	736	94.43%
	D	1	4	0	779	0	0	0	7	0	0	0	0	0	6	4	1	1	1	1	0	0	0	0	0	0	0	805	96.77%
	E	0	3	4	0	720	4	12	0	0	0	2	5	0	0	0	1	2	0	2	1	1	0	0	4	0	7	768	93.75%
	F	0	2	0	3	1	725	0	1	1	1	0	0	0	3	0	21	0	0	1	11	0	1	2	1	1	0	775	93.55%
	G	0	7	3	10	9	0	725	3	0	0	0	0	5	0	2	1	2	1	1	0	0	1	1	1	1	0	773	93.79%
	H	0	16	0	18	0	2	6	623	0	0	26	1	2	5	3	4	3	17	0	1	5	1	0	0	0	1	734	84.88%
	I	0	0	0	2	1	1	0	0	722	25	0	2	0	1	0	0	0	0	0	0	0	0	0	0	1	0	755	95.63%
	J	0	1	0	2	0	3	0	2	31	693	0	1	0	2	1	0	2	0	1	0	4	0	0	3	0	1	747	92.77%
	K	0	5	1	5	8	0	0	15	0	0	657	2	0	0	0	1	0	12	0	1	4	0	1	27	0	0	739	88.90%
	L	0	0	0	0	4	0	7	0	0	1	2	742	0	0	0	0	1	3	0	0	0	0	0	1	0	0	761	97.50%
	M	1	0	0	0	0	0	2	0	0	0	0	0	775	2	0	0	0	0	0	0	3	4	5	0	0	0	792	97.85%
	N	1	2	0	11	0	0	0	3	0	0	0	2	4	743	4	0	1	5	0	2	0	4	1	0	0	0	783	94.89%
	O	0	1	3	12	0	0	1	0	0	0	0	0	0	5	711	0	10	1	0	1	5	0	3	0	0	0	753	94.42%
	P	0	1	0	0	0	40	0	2	0	0	0	2	1	0	0	750	1	1	0	2	0	0	0	0	3	0	803	93.40%
	Q	0	0	0	1	0	0	0	0	0	0	0	1	0	0	23	8	743	4	0	0	2	0	0	1	0	0	783	94.89%
	R	0	24	0	6	1	1	1	3	0	0	6	4	1	17	0	1	2	688	0	1	1	1	0	0	0	0	758	90.77%
	S	0	2	0	3	6	3	1	1	1	1	0	0	0	0	0	0	1	3	723	1	0	0	0	0	0	2	748	96.66%
	T	0	1	1	1	0	2	0	0	0	1	2	0	0	0	0	0	0	0	0	775	0	0	0	0	13	0	796	97.36%
	U	2	0	0	0	0	0	0	3	0	0	0	0	2	1	1	0	0	0	0	0	804	0	0	0	0	0	813	98.89%
	V	0	8	1	0	0	2	1	0	0	0	0	0	5	2	1	2	0	0	0	1	1	733	4	0	3	0	764	95.94%
	W	0	0	0	0	0	0	0	0	0	0	0	0	4	0	4	0	0	1	0	0	7	3	732	0	1	0	752	97.34%
	X	0	0	1	3	5	0	0	0	0	0	13	3	0	0	2	1	0	1	1	2	1	0	0	753	1	0	787	95.68%
	Y	2	1	0	0	0	0	0	0	0	0	0	0	1	0	0	0	1	0	0	11	2	2	0	1	765	0	786	97.33%
	Z	0	0	0	0	4	1	0	0	0	0	0	1	0	0	0	0	8	0	3	1	0	0	0	0	0	716	734	97.55%
Total		787	798	711	865	772	786	767	665	755	722	709	768	804	789	764	791	780	748	734	813	849	758	752	794	791	728	20000	94.91%

Letter * Predicted Value for Letter Crosstabulation(K=5)																													
Count		Predicted Value for Letter																				Total							
	A	B	C	D	E	F	G	H	I	J	K	L	M	N	O	P	Q	R	S	T	U	V	W	X	Y	Z			
Letter	A	780	0	2	0	0	0	0	0	0	0	1	1	2	0	0	0	2	0	0	0	0	0	0	1	0	789	98.86%	
	B	0	717	0	7	4	3	0	5	0	0	1	0	4	1	0	0	0	13	1	0	3	7	0	0	0	766	93.60%	
	C	0	0	694	0	5	0	11	0	0	0	0	2	0	0	14	0	2	0	0	1	3	1	1	2	0	736	94.29%	
	D	0	4	0	782	1	0	0	6	0	0	0	1	4	1	0	1	2	2	0	0	0	0	1	0	0	805	97.14%	
	E	0	3	4	0	729	1	11	0	0	0	3	1	0	0	0	3	0	2	1	0	0	0	2	0	8	768	94.92%	
	F	0	2	0	3	0	716	0	0	0	1	0	0	0	4	0	28	0	0	0	17	0	2	0	1	1	0	775	92.39%
	G	0	2	3	11	11	1	723	5	0	0	0	2	0	6	1	3	1	1	0	0	1	1	1	1	0	0	773	93.63%
	H	0	19	1	19	0	0	4	611	0	0	27	0	1	2	9	2	2	29	0	0	3	1	0	3	1	0	734	83.24%
	I	0	1	0	3	0	5	0	0	722	22	1	1	0	0	0	0	0	0	0	0	0	0	0	0	0	0	755	95.63%
	J	1	0	0	0	0	2	0	1	36	692	0	0	0	1	4	0	1	0	0	0	4	0	0	3	0	2	747	92.64%
	K	0	5	0	4	16	0	3	13	0	0	655	2	0	0	0	0	0	19	1	2	3	0	0	16	0	0	739	88.63%
	L	0	0	0	0	3	0	3	2	0	1	0	743	0	0	0	0	1	6	0	0	0	0	2	0	0	761	97.63%	
	M	0	3	0	0	0	0	4	0	0	0	0	0	778	2	0	0	0	0	0	1	1	3	0	0	0	792	98.23%	
	N	0	1	0	13	0	0	0	4	0	0	0	1	4	738	6	0	0	11	0	0	0	4	1	0	0	0	783	94.25%
	O	0	1	3	15	1	0	0	0	0	0	0	1	4	714	0	8	0	0	0	3	1	2	0	0	0	0	753	94.82%
	P	0	2	0	3	2	42	1	3	0	0	0	1	0	0	745	1	2	0	0	0	0	0	0	1	0	803	92.78%	
	Q	0	0	0	2	2	0	0	0	0	0	0	0	0	31	5	741	1	0	0	0	0	0	0	1	0	783	94.64%	
	R	0	16	0	4	2	0	0	6	0	0	3	5	1	3	0	0	1	714	0	0	0	3	0	0	0	0	758	94.20%
	S	0	7	0	3	6	5	0	0	0	0	0	1	0	0	0	1	6	715	2	0	0	0	0	0	2	748	95.59%	
	T	0	2	2	3	0	2	0	2	1	0	0	0	0	0	0	1	1	0	768	0	2	0	1	11	0	796	96.48%	
	U	2	0	1	1	0	0	0	5	0	0	1	0	3	0	1	0	0	0	0	0	798	1	0	0	0	813	98.15%	
	V	0	10	0	1	0	3	1	0	0	0	0	0	4	0	1	2	0	0	0	0	2	733	5	0	2	764	95.94%	
	W	0	1	0	0	0	1	0	0	0	0	1	0	9	0	5	0	0	0	0	5	3	727	0	0	0	752	96.68%	
	X	1	2	1	1	5	0	0	0	0	0	17	0	0	0	0	2	2	3	3	1	0	0	749	0	0	787	95.17%	
	Y	4	0	0	0	0	1	0	0	0	0	0	0	0	1	0	3	0	0	12	1	7	0	1	756	0	786	96.18%	
	Z	0	0	0	0	2	0	0	0	0	1	0	0	0	0	0	8	0	1	2	0	0	0	1	0	719	734	97.96%	
Total		788	798	711	875	789	781	762	663	759	717	709	757	811	761	792	786	776	809	726	808	827	767	740	783	774	731	20000	94.75%

Letter * Predicted Value for Letter Crosstabulation (K=7)																															
Count		Predicted Value for Letter																										Total			
	A	B	C	D	E	F	G	H	I	J	K	L	M	N	O	P	Q	R	S	T	U	V	W	X	Y	Z					
Letter	A	781	0	1	0	0	0	0	1	0	0	0	2	0	0	0	0	1	0	0	2	0	0	0	0	1	0	789	98.99%		
	B	0	710	0	2	7	1	0	8	0	0	1	0	1	1	0	0	0	21	4	0	1	6	0	3	0	766	92.69%			
	C	1	0	696	0	3	0	11	1	0	0	1	2	0	1	10	0	1	1	0	1	2	1	1	2	1	0	736	94.57%		
	D	1	4	0	776	0	0	1	9	1	1	0	0	1	5	2	0	1	1	1	0	0	0	0	1	0	0	805	96.40%		
	E	0	2	7	0	712	2	13	1	1	0	4	0	1	0	0	5	0	5	0	2	0	2	0	6	0	10	768	92.71%		
	F	2	0	0	5	0	694	1	1	14	2	0	0	1	5	1	29	0	1	1	15	0	0	3	0	0	0	775	89.55%		
	G	0	7	3	9	13	0	716	2	0	0	0	0	1	0	6	0	2	1	3	0	0	2	2	5	1	0	773	92.63%		
	H	0	8	1	28	0	1	5	616	0	0	21	0	1	1	10	2	1	25	2	1	1	1	0	2	5	2	734	93.92%		
	I	0	0	0	2	0	2	0	0	727	21	1	1	0	0	0	0	0	0	0	0	0	0	0	1	0	0	755	96.29%		
	J	1	0	0	1	1	1	0	2	44	682	0	1	0	1	3	1	1	1	0	0	3	0	0	3	0	1	747	91.30%		
	K	0	3	0	4	14	0	5	27	0	0	637	2	0	0	0	0	0	20	1	0	3	0	0	23	0	0	739	86.20%		
	L	0	0	0	0	4	0	5	1	1	1	0	736	0	0	0	0	1	7	1	0	1	0	0	3	0	0	761	96.71%		
	M	1	3	0	1	0	0	4	0	0	0	2	0	762	9	2	1	0	0	0	0	0	4	3	0	0	0	792	96.21%		
	N	2	1	0	7	0	0	0	4	0	0	0	0	6	732	9	0	0	15	0	0	1	4	2	0	0	0	783	93.49%		
	O	0	1	4	13	0	0	2	0	0	0	0	0	2	718	0	6	1	0	0	2	1	2	1	0	0	0	753	95.35%		
	P	0	3	0	1	1	34	0	3	0	0	0	1	0	0	2	752	1	1	0	0	0	0	0	0	4	0	803	93.65%		
	Q	0	0	0	2	0	0	2	0	0	0	0	0	0	20	3	751	3	0	0	0	0	0	0	0	2	0	783	95.91%		
	R	0	19	0	5	0	0	0	9	0	1	7	2	1	1	0	0	0	710	0	0	0	3	0	0	0	0	758	93.67%		
	S	0	3	0	2	6	0	0	1	0	0	0	1	0	0	0	0	1	1	727	2	1	0	0	0	1	0	2	748	97.19%	
	T	0	1	1	4	1	2	0	2	1	0	0	0	0	0	1	1	1	1	0	760	0	1	0	3	15	1	796	95.48%		
	U	4	0	0	1	0	0	0	7	0	0	0	0	3	0	0	0	0	0	0	0	797	1	0	0	0	0	813	98.03%		
	V	1	12	0	0	0	3	1	0	0	0	0	0	1	1	3	3	0	2	0	1	2	729	4	0	1	0	764	95.42%		
	W	0	1	0	0	0	0	1	0	0	0	0	0	3	0	5	0	0	0	0	0	3	2	736	0	1	0	752	97.87%		
	X	1	2	0	3	8	0	0	0	0	0	8	1	0	0	0	0	2	3	5	0	2	0	0	752	0	0	787	95.55%		
	Y	0	1	0	0	0	0	0	0	0	0	0	0	0	0	0	2	1	0	0	7	2	5	2	1	765	0	786	97.33%		
	Z	0	0	0	0	2	0	0	0	0	1	0	0	0	0	0	0	10	1	3	0	0	0	0	2	0	715	734	97.41%		
Total		795	781	713	866	772	740	767	695	789	709	682	747	784	759	792	794	785	817	750	787	823	762	755	809	796	731	20000	94.40%		

The Following is the binned data result:

Letter * Predicted Value for Letter Crosstabulation(K=1, Binned)																													
Count																													
		Predicted Value for Letter																										Total	
Letter	A	B	C	D	E	F	G	H	I	J	K	L	M	N	O	P	Q	R	S	T	U	V	W	X	Y	Z			
A	730	3	0	0	1	0	1	3	0	5	2	15	5	7	2	0	2	1	3	0	3	0	0	3	1	2	789	92.52%	
B	2	528	0	27	15	5	11	18	2	3	10	3	3	5	3	10	11	39	27	7	5	4	5	14	4	5	766	88.93%	
C	1	1	614	0	17	6	26	3	0	0	20	1	0	2	11	4	3	5	4	4	7	0	0	4	1	2	736	83.42%	
D	1	28	3	632	5	5	6	19	3	1	1	1	1	10	23	8	5	13	7	5	6	3	2	7	3	7	805	78.51%	
E	1	17	15	2	558	0	31	3	1	2	10	5	1	1	1	0	5	7	31	4	0	2	0	21	0	50	768	72.66%	
F	2	4	4	9	3	609	0	4	4	13	3	5	0	7	0	48	0	3	8	30	2	5	2	3	6	1	775	78.58%	
G	1	16	13	8	23	1	610	10	2	0	8	2	3	3	11	4	25	13	4	4	4	3	1	2	0	2	773	78.91%	
H	2	24	1	19	6	3	11	482	1	2	31	2	2	18	43	7	5	40	6	1	1	1	4	15	3	4	734	65.67%	
I	1	1	0	4	2	2	3	0	631	41	0	6	0	0	44	1	5	1	1	1	0	0	0	6	1	4	755	83.58%	
J	3	7	1	5	2	10	1	3	55	612	1	4	0	1	6	3	3	5	10	1	1	0	0	7	2	4	747	81.93%	
K	3	18	15	4	40	2	15	34	1	3	493	4	4	7	5	3	1	25	7	3	1	3	1	40	0	7	739	66.71%	
L	12	1	1	0	3	4	12	2	5	1	0	692	0	1	0	4	3	8	2	2	0	0	0	5	0	3	761	90.93%	
M	7	6	3	1	2	0	3	4	0	0	7	0	712	13	3	0	2	3	0	1	5	4	15	1	0	0	792	89.90%	
N	5	6	1	14	1	3	0	25	0	3	3	0	2	667	11	4	3	11	1	1	4	4	10	3	1	0	783	85.19%	
O	0	3	12	22	2	1	10	30	2	5	0	2	3	8	579	1	42	5	4	0	8	4	8	1	0	1	753	76.89%	
P	0	13	0	15	2	55	6	1	4	1	1	4	0	6	3	649	4	6	6	4	2	2	3	1	15	0	803	80.82%	
Q	1	7	1	12	5	1	26	3	3	2	0	1	2	4	45	2	638	4	8	1	4	1	1	6	2	3	783	81.48%	
R	6	64	4	17	6	3	19	17	4	2	21	9	8	9	3	4	6	525	5	6	0	3	0	15	1	1	758	69.26%	
S	1	27	0	8	35	8	5	6	7	9	3	0	3	0	2	4	9	7	533	3	1	3	1	14	0	59	748	71.26%	
T	0	6	6	3	7	39	4	5	2	1	1	1	0	3	1	3	3	4	4	630	5	8	1	1	52	6	796	79.15%	
U	3	1	21	3	0	5	4	5	0	2	3	0	12	7	10	0	6	1	1	6	709	6	5	0	3	0	813	87.21%	
V	0	5	0	3	1	6	4	3	0	0	4	0	2	3	3	5	0	2	2	8	4	609	16	0	84	0	764	79.71%	
W	0	3	0	1	0	0	0	4	0	0	1	0	14	2	3	0	0	1	2	0	5	13	703	0	0	0	752	93.48%	
X	2	31	2	5	23	0	2	14	11	0	27	5	0	2	6	0	9	9	17	0	2	0	0	597	0	23	787	75.86%	
Y	3	5	1	1	1	5	0	8	2	1	0	1	0	1	1	8	4	2	3	54	4	49	4	2	624	2	786	79.39%	
Z	0	4	2	5	35	1	2	2	6	5	0	4	0	2	0	0	6	0	46	6	0	0	0	11	0	597	734	81.34%	
Total	787	829	720	820	795	774	812	708	746	714	650	767	777	789	819	772	800	740	742	782	783	727	782	779	803	783	20000	79.74%	

Letter * Predicted Value for Letter Crosstabulation(K=3,Binned)																													
Count																													
		Predicted Value for Letter																										Total	
		A	B	C	D	E	F	G	H	I	J	K	L	M	N	O	P	Q	R	S	T	U	V	W	X	Y	Z		
Letter	A	738	4	0	4	1	0	2	0	3	2	1	6	2	5	2	3	0	3	3	0	9	0	0	0	0	1	789	93.54%
	B	6	515	0	52	10	4	19	8	7	1	3	2	5	2	6	9	5	34	30	6	5	13	0	20	3	1	766	67.23%
	C	3	2	565	4	17	6	50	1	4	0	17	0	0	0	19	5	4	11	3	6	12	1	0	2	2	2	736	76.77%
	D	0	21	2	692	4	1	5	8	3	6	1	0	1	7	19	5	5	6	4	1	4	2	0	4	2	2	805	85.96%
	E	3	21	10	8	513	4	35	2	0	0	17	4	0	1	1	1	4	13	31	8	0	2	0	29	1	60	768	66.80%
	F	2	10	1	11	3	580	0	8	8	3	0	2	1	7	0	54	1	8	4	44	4	10	1	5	7	1	775	74.84%
	G	2	6	19	26	19	1	604	2	1	0	10	2	3	1	11	6	15	17	3	2	8	3	4	5	1	2	773	78.14%
	H	3	42	1	39	6	0	16	441	24	1	15	3	1	10	30	14	4	29	4	5	12	2	2	21	6	3	734	60.08%
	I	1	6	1	7	0	2	2	0	689	27	1	2	0	0	3	1	0	2	4	2	0	1	0	2	1	1	755	91.26%
	J	4	8	1	6	1	11	0	4	57	599	0	4	0	2	7	7	3	11	6	0	1	0	0	7	4	4	747	80.19%
	K	4	19	16	6	24	5	20	28	2	4	456	5	8	10	5	6	2	27	9	4	3	6	2	58	1	9	739	61.71%
	L	19	7	0	3	1	2	14	0	10	1	1	679	0	0	1	3	2	7	0	3	3	0	0	1	2	2	761	89.22%
	M	11	4	1	5	0	1	7	0	0	0	6	0	723	6	1	0	1	1	1	2	11	4	7	0	0	0	792	91.29%
	N	11	10	1	22	3	4	1	18	2	2	2	1	8	639	17	3	2	8	1	4	7	5	10	2	0	0	783	81.61%
	O	0	1	5	27	1	0	31	10	40	5	0	0	2	1	570	3	33	3	4	1	8	4	2	1	1	0	753	75.70%
	P	1	13	0	18	1	63	5	3	4	0	1	1	0	2	4	652	3	6	5	4	2	2	1	0	11	1	803	81.20%
	Q	1	14	2	14	2	0	36	4	11	2	0	1	2	2	50	11	610	5	2	1	4	2	0	3	2	2	783	77.91%
	R	10	69	3	36	8	2	19	11	4	3	10	2	9	8	9	7	2	512	3	6	1	6	0	15	3	0	758	67.55%
	S	2	29	1	20	34	13	6	1	10	5	2	0	4	1	10	2	2	14	498	8	2	6	0	15	4	59	748	66.58%
	T	2	7	1	5	4	27	2	3	0	0	0	0	1	3	2	6	2	5	0	635	5	9	0	1	69	7	796	79.77%
	U	7	2	13	5	0	3	2	4	0	0	2	0	6	3	3	5	7	0	0	2	738	5	1	0	5	0	813	90.77%
	V	1	9	0	6	0	3	6	1	0	0	0	0	1	6	5	10	0	2	2	4	3	613	11	0	81	0	764	80.24%
	W	1	3	0	1	0	0	3	1	0	0	0	0	20	2	5	2	0	1	0	0	10	14	688	0	1	0	752	91.49%
	X	3	17	2	12	18	2	5	3	21	3	28	1	2	2	9	3	8	10	16	3	4	1	0	594	1	19	787	75.48%
	Y	5	3	0	0	0	6	2	1	2	2	2	0	0	0	0	18	2	0	1	59	4	79	2	1	594	3	786	75.57%
	Z	1	10	0	7	28	1	3	0	10	3	1	1	0	0	1	0	7	1	75	6	1	1	0	17	3	557	734	75.89%
Total		841	852	645	###	698	741	895	562	912	669	576	716	799	720	790	836	724	736	709	816	861	791	731	803	805	736	20000	78.34%

Letter * Predicted Value for Letter Crosstabulation (K=5,Binned)																													
Count																													
		Predicted Value for Letter																										Total	
		A	B	C	D	E	F	G	H	I	J	K	L	M	N	O	P	Q	R	S	T	U	V	W	X	Y	Z		
Letter	A	727	3	0	3	2	2	0	2	3	2	2	4	7	2	2	1	1	6	3	1	8	0	0	5	0	3	789	92.14%
	B	0	477	0	44	18	9	9	20	2	4	5	1	8	2	2	5	6	56	44	2	7	4	0	31	4	6	766	62.27%
	C	1	0	587	3	26	3	25	4	1	2	20	1	1	2	12	2	5	4	6	5	19	1	0	3	1	2	736	79.76%
	D	0	17	5	686	3	1	2	15	0	4	0	0	1	7	14	8	7	14	5	0	5	0	0	7	2	2	805	85.22%
	E	0	6	9	3	577	4	28	3	0	1	20	6	0	1	3	1	6	5	26	0	1	0	0	34	0	34	768	75.13%
	F	1	1	2	12	4	585	0	4	5	20	1	9	0	11	0	40	0	2	10	58	2	2	1	4	1	0	775	75.48%
	G	3	7	24	21	25	1	595	4	0	2	4	2	2	3	8	8	25	17	1	1	5	3	0	12	0	0	773	76.97%
	H	0	22	0	33	9	0	6	505	0	1	24	3	1	12	16	10	6	32	10	2	6	1	1	27	5	2	734	68.80%
	I	0	4	1	5	3	4	2	44	616	46	0	4	0	0	5	4	3	5	3	0	0	0	4	0	2	755	81.59%	
	J	0	2	1	14	1	12	0	4	39	627	0	5	0	1	3	2	2	3	9	1	4	0	0	13	0	4	747	83.94%
	K	3	8	12	9	35	5	15	43	0	2	470	2	10	9	3	3	3	29	6	2	2	2	1	58	2	5	739	63.60%
	L	22	1	0	3	3	1	9	2	1	2	2	689	0	0	0	4	4	5	2	1	2	0	0	6	2	0	761	90.54%
	M	9	3	3	6	1	3	1	5	0	0	8	0	708	16	3	1	0	3	1	0	7	3	9	2	0	0	792	89.39%
	N	8	8	1	25	4	0	0	23	2	2	2	1	13	631	15	2	3	11	1	0	10	4	14	3	0	0	783	80.59%
	O	0	1	11	26	2	0	9	51	5	9	0	2	1	0	567	4	29	3	7	0	6	10	7	3	0	0	753	75.30%
	P	0	12	0	18	0	64	4	2	1	3	0	3	0	3	3	645	5	3	8	9	0	2	2	5	11	0	803	80.32%
	Q	0	5	3	14	4	0	21	2	8	3	1	1	0	0	29	5	654	4	7	0	5	5	1	5	4	2	783	83.52%
	R	4	48	6	31	12	5	13	25	3	7	8	1	7	10	6	0	8	529	6	3	2	1	0	22	1	0	758	69.79%
	S	1	26	1	10	58	10	1	4	3	8	2	0	0	0	4	6	6	9	522	3	3	0	0	13	2	56	748	69.79%
	T	1	9	2	5	5	23	2	7	0	1	1	1	0	1	3	2	5	4	3	637	7	11	2	3	54	7	796	80.03%
	U	2	2	11	6	1	7	2	5	0	2	3	0	3	4	10	1	8	0	0	1	726	7	4	1	7	0	813	89.30%
	V	1	6	0	6	0	8	3	1	0	1	1	0	6	5	0	9	0	2	1	8	3	625	10	1	67	0	764	81.81%
	W	0	3	0	1	0	1	0	2	0	0	0	0	13	1	2	2	1	5	0	2	10	20	685	0	4	0	752	91.09%
	X	0	4	0	3	33	1	1	6	14	3	23	3	2	1	3	2	5	8	15	2	5	0	0	637	1	15	787	80.94%
	Y	2	3	0	1	0	7	0	2	1	7	2	2	1	0	0	21	6	1	1	51	4	60	1	4	608	1	786	77.35%
	Z	0	4	0	3	63	2	1	3	3	6	0	3	0	0	1	0	7	7	62	1	2	0	0	10	1	555	734	75.61%
Total		785	682	679	991	889	758	749	788	707	765	599	743	784	722	709	789	806	765	761	793	851	761	738	913	777	696	20000	79.24%

Letter * Predicted Value for Letter Crosstabulation(K=7, Binned)																													
Count		Predicted Value for Letter																										Total	
		A	B	C	D	E	F	G	H	I	J	K	L	M	N	O	P	Q	R	S	T	U	V	W	X	Y	Z		
Letter	A	741	2	0	1	0	0	1	0	1	0	1	9	8	5	1	1	0	5	4	1	1	0	0	1	5	1	789	93.92%
	B	2	566	0	21	6	15	3	3	2	1	3	1	3	3	4	1	12	51	18	5	6	5	2	22	7	4	766	73.89%
	C	1	4	561	6	22	7	26	3	0	1	13	1	0	0	19	1	9	9	8	9	21	3	3	3	4	2	736	76.22%
	D	0	26	2	663	2	2	2	12	1	3	4	0	2	14	22	5	8	14	5	4	5	0	0	4	4	1	805	82.36%
	E	1	19	13	0	506	3	22	7	0	1	31	4	1	1	2	2	9	6	34	4	0	0	2	39	1	60	768	65.89%
	F	1	9	2	9	2	593	0	2	9	4	1	2	0	8	0	43	1	6	7	55	4	4	3	3	6	1	775	76.52%
	G	3	15	17	12	18	2	557	8	0	1	9	0	3	4	17	8	33	28	2	4	9	6	3	10	3	1	773	72.06%
	H	3	31	1	32	5	4	7	475	25	0	19	3	3	17	24	12	6	27	6	1	9	0	1	17	3	3	734	64.71%
	I	2	11	1	2	1	6	1	0	680	26	2	3	0	0	0	0	3	1	6	2	0	0	0	2	2	4	755	90.07%
	J	2	8	0	8	1	12	2	3	54	598	1	6	0	6	2	2	5	6	12	1	2	0	0	9	5	2	747	80.05%
	K	3	25	4	7	16	4	15	36	3	1	478	5	2	16	3	5	3	32	8	3	3	4	0	48	3	12	739	64.68%
	L	28	1	0	1	6	1	5	2	4	1	1	674	0	1	0	3	4	11	3	1	1	0	0	10	3	0	761	88.57%
	M	12	10	1	2	0	1	1	1	0	0	7	0	711	10	1	1	1	8	1	2	11	2	9	0	0	0	792	89.77%
	N	7	9	0	34	0	0	4	15	2	1	2	0	13	642	16	2	1	7	0	0	10	5	11	0	2	0	783	81.99%
	O	0	1	9	19	0	1	6	10	40	2	0	1	2	2	582	3	41	6	6	1	6	6	7	1	1	0	753	77.29%
	P	1	15	0	23	1	75	8	4	2	3	2	4	0	3	3	600	5	6	8	11	0	2	1	4	22	0	803	74.72%
	Q	1	9	2	14	5	0	14	4	4	1	3	1	1	1	33	4	656	3	11	1	3	2	2	3	4	1	783	83.78%
	R	1	74	0	27	6	2	5	14	6	5	20	1	10	6	3	3	6	531	7	5	1	0	2	22	1	0	758	70.05%
	S	3	41	1	8	30	10	5	1	4	7	0	0	3	3	3	4	3	13	504	5	4	1	0	14	3	78	748	67.38%
	T	0	9	1	3	6	25	1	5	0	0	1	2	1	3	2	3	3	12	3	629	6	6	1	5	63	6	796	79.02%
	U	4	0	5	5	0	4	1	2	1	1	0	0	7	7	12	1	5	1	0	3	739	7	3	1	4	0	813	90.90%
	V	1	13	0	3	0	9	1	1	2	0	2	0	0	6	0	6	1	3	3	13	4	604	19	0	73	0	764	79.06%
	W	0	8	0	1	0	0	0	2	0	0	0	0	15	2	3	0	0	5	1	0	9	15	688	0	3	0	752	91.49%
	X	3	11	0	4	13	2	1	7	15	4	25	3	0	1	3	1	7	8	24	2	7	0	0	616	1	29	787	78.27%
	Y	1	3	0	0	1	5	1	4	1	0	0	1	2	1	2	9	2	1	2	52	4	58	2	3	630	1	786	80.15%
	Z	0	12	1	6	46	1	1	0	6	4	0	4	0	0	2	0	7	9	44	4	1	0	0	8	3	575	734	78.34%
Total		821	932	621	911	693	784	690	621	862	665	625	725	787	762	759	720	831	809	727	818	866	730	759	845	856	781	20000	78.89%

IS467
Homework#4
Date: 06-03-2017

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Question 2-c:

Ans:

Based on the above result, the result from the K-nearest neighbor seems have a more accurate result than the decision tree. However, The running time of K-nearest neighbor is a lot longer than the Decision. In my opinion, the accuracy measure may be a good idea for comparing the Training and Testing data for further verification. In the SPSS, there is not a lot of parameters setting available for K-nearest neighbor analysis. On the contrary, there is more adjustment available (Depth, Parents node and Child nodes etc.) on the decision tree algorithms. Also, I would prefer to use Decision Tree algorithms for the future analysis.

Question3

a.

- 1) In k-means how are the cluster centers calculated?

Ans:

First of all, clusters are defined by their centers, the following is the steps to process calculation.

- First, the algorithm arbitrarily choose K object as initial cluster center.
- And then assign each objects to most similar center.
- Then clusters update their cluster Centroids (i.e. Mean point)
- Then the objects will be assigned to the most similar center by the distance function
- Repeat step 3 and 4 until NO change (=No better result) of centroids

- 2) Name two similarity measures (or distance functions) and what type of data you would use them for.

Ans:

- Minkowski Distance – which is good for two p-dimensional data object and q is a positive integer
- Manhattan distance – which is good for 2-dimensional data, especially only have it (x, y) coordinates along the axis and q=1
- Euclidean distance – which is good for 2-dimensional data and with q=2.

- 3) Perform k-means clustering:

i & ii Report the final cluster centers and the number of elements in each clustering

k=3:

Final Cluster Centers			
	Cluster		
	1	2	3
Area	18.72	11.96	14.65
Perimeter	16.30	13.27	14.46
Compactness	.89	.85	.88
Length_kernel	6.21	5.23	5.56
width_kernel	3.72	2.87	3.28
asymmetry_coef	3.60	4.76	2.65
length_ker_groove	6.07	5.09	5.19

Number of Cases in each Cluster

Cluster	1	61.000
	2	77.000
	3	72.000
Valid		210.000
Missing		.000

k=4:

Final Cluster Centers				
	Cluster			
	1	2	3	4
Area	11.94	14.42	17.75	19.52
Perimeter	13.27	14.35	15.88	16.65
Compactness	.85	.88	.88	.88
Length_kernel	5.23	5.52	6.05	6.35
width_kernel	2.87	3.25	3.61	3.81
asymmetry_coef	4.80	2.59	3.16	4.16
length_ker_groove	5.10	5.13	5.92	6.18

Number of Cases in each Cluster

Cluster	1	75.000
	2	67.000
	3	40.000
	4	28.000
Valid		210.000
Missing		.000

k=5

Final Cluster Centers					
	Cluster				
	1	2	3	4	5
Area	16.56	14.69	19.15	12.09	11.98
Perimeter	15.39	14.47	16.47	13.31	13.29
Compactness	.88	.88	.89	.86	.85
Length_kernel	5.89	5.57	6.27	5.22	5.24
width_kernel	3.48	3.29	3.77	2.90	2.88
asymmetry_coef	4.11	2.41	3.46	3.34	5.67
length_ker_groove	5.73	5.16	6.13	5.01	5.12

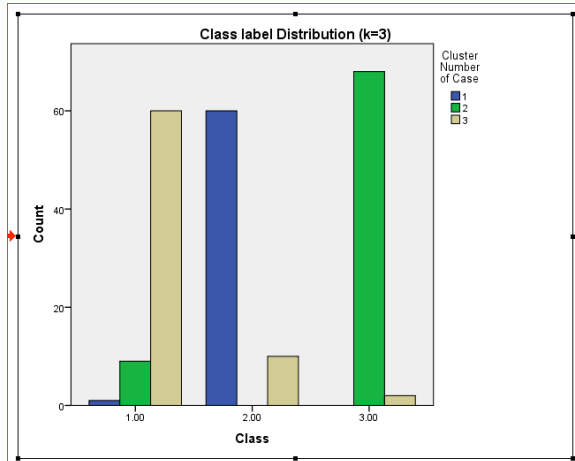
Number of Cases in each Cluster		
Cluster	1	25.000
	2	51.000
	3	48.000
	4	44.000
	5	42.000
Valid		210.000
Missing		.000

k=6

Final Cluster Centers						
	Cluster					
	1	2	3	4	5	6
Area	11.83	14.24	16.41	18.95	12.32	19.58
Perimeter	13.22	14.26	15.32	16.39	13.42	16.65
Compactness	.85	.88	.88	.89	.86	.89
Length_kernel	5.22	5.49	5.86	6.25	5.27	6.32
width_kernel	2.84	3.23	3.46	3.74	2.95	3.84
asymmetry_coef	4.17	2.32	3.85	2.72	6.34	5.08
length_ker_groove	5.08	5.06	5.69	6.12	5.12	6.14

Number of Cases in each Cluster		
Cluster	1	56.000
	2	54.000
	3	31.000
	4	33.000
	5	21.000
	6	15.000
Valid		210.000
Missing		.000

iii. Report the class distribution within each cluster
(use crosstab between labels and cluster membership)
k=3

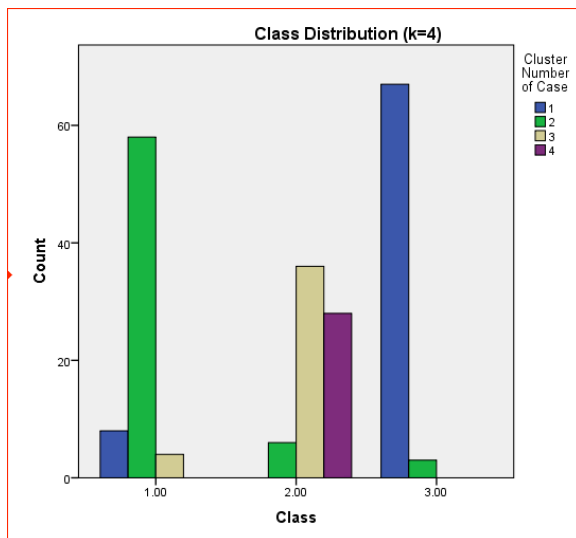


Class * Cluster Number of Case Crosstabulation

Count

		Cluster Number of Case			
		1	2	3	Total
Class	1.00	1	9	60	70
	2.00	60	0	10	70
	3.00	0	68	2	70
Total		61	77	72	210

k=4

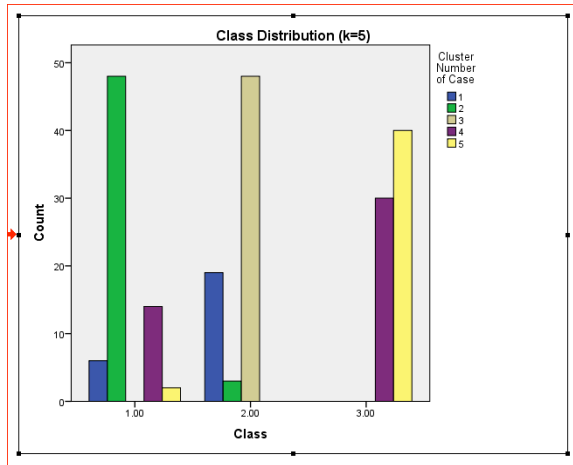


Class * Cluster Number of Case Crosstabulation

Count

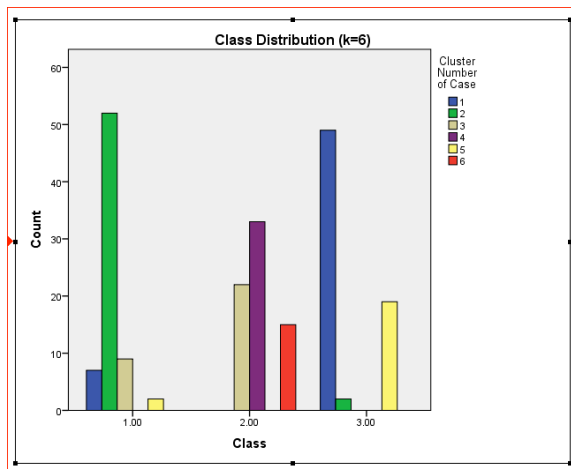
		Cluster Number of Case				
		1	2	3	4	Total
Class	1.00	8	58	4	0	70
	2.00	0	6	36	28	70
	3.00	67	3	0	0	70
Total		75	67	40	28	210

K=5



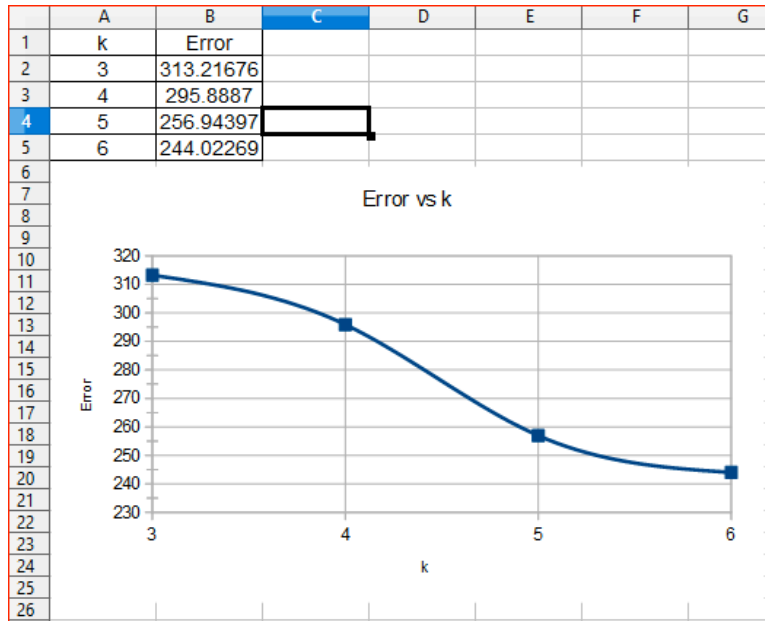
Class * Cluster Number of Case Crosstabulation							
Count		Cluster Number of Case					Total
		1	2	3	4	5	
Class	1.00	6	48	0	14	2	70
	2.00	19	3	48	0	0	70
	3.00	0	0	0	30	40	70
Total		25	51	48	44	42	210

K=6



Class * Cluster Number of Case Crosstabulation							
Count		Cluster Number of Case					
		1	2	3	4	5	6
Class	1.00	7	52	9	0	2	0
	2.00	0	0	22	33	0	15
	3.00	49	2	0	0	19	0
Total		56	54	31	33	21	15

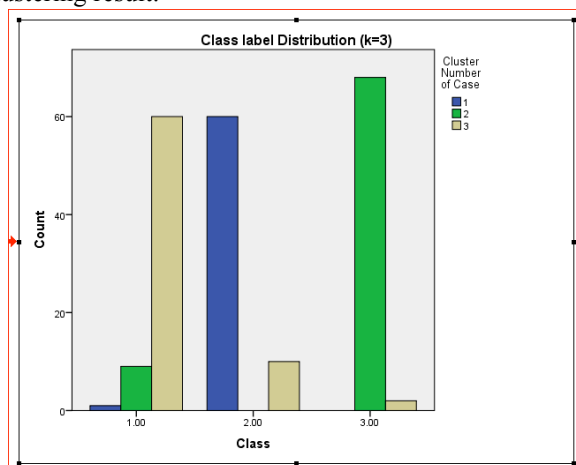
Question3a
Part4)



Based on the above , I would choose the clustering with k=5 because the curve seems touch the knee after the k=5 section. In other words, there is not much improvements on the accuracy even increasing the number of clusters to the

Question3a
Part 5:

Ans: According to the part 3a iii), I would choose clusters with k=3 shown below. The reason is that the result most likely matches to the pattern of class labels originally provided from the Table even though their labels' number NOT named the same. Although the clustering is not perfectly classified, they majority of clusters matches to the classes labels, especially the Cluster #2 (Green bar) matches to the class label #3 more than 60 counts. So that I would conclude that k=3 has the best clustering result.

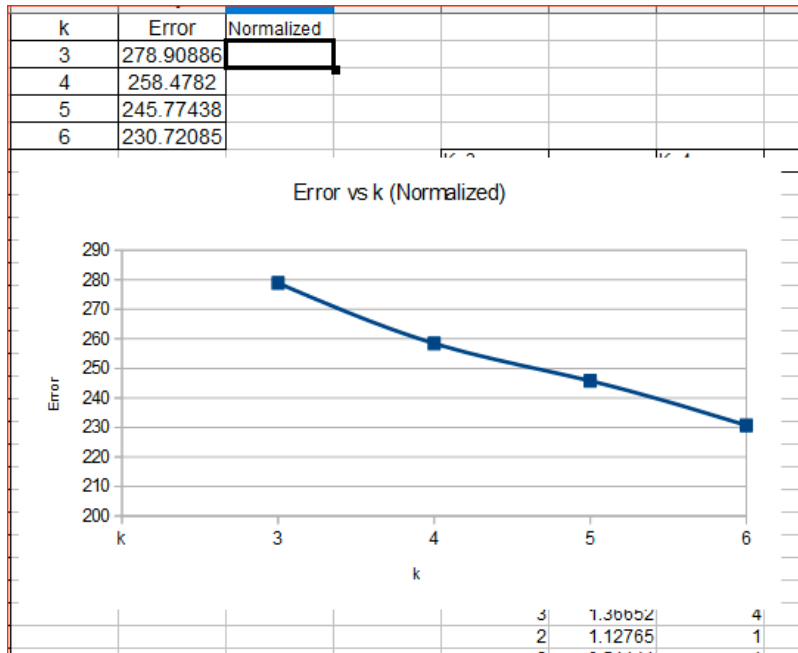


Class * Cluster Number of Case Crosstabulation				
Count		Cluster Number of Case		
		1	2	3
Class	1.00	1	9	60
	2.00	60	0	10
	3.00	0	68	2
Total		61	77	72

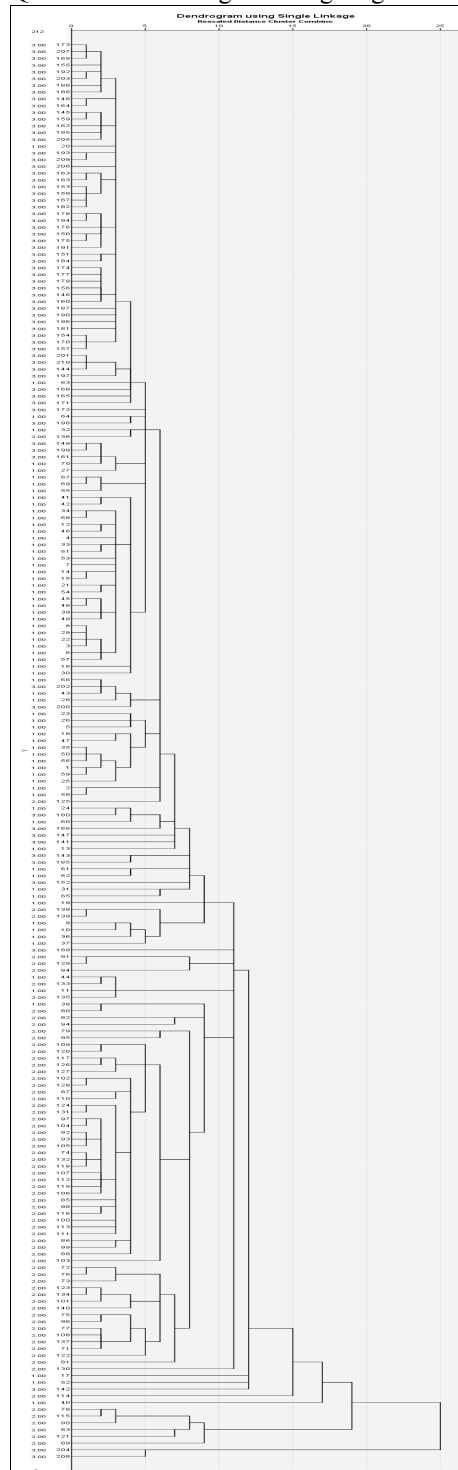
Question 3a

Part 6:

Based on the following plot for normalized data, I would say that the normalized data gives out a more accurate result. When we observe the plot, we kind of NOT able to see the knee or elbow section which means that more clusters ($k > 7$) might be able to add for analysis to increase the accuracy. In part 4, the cluster's knee is at about $k=5$ with error=256. In this result, the error probably lower than 230 with $k>6$. The reason behind that is probably some of the attributes (e.g. Area, Perimeter) are too large comparing the other attributes in the dataset. And The distance function is sensitive to the data variables with very wide range because the equation involves square of the subtract values.



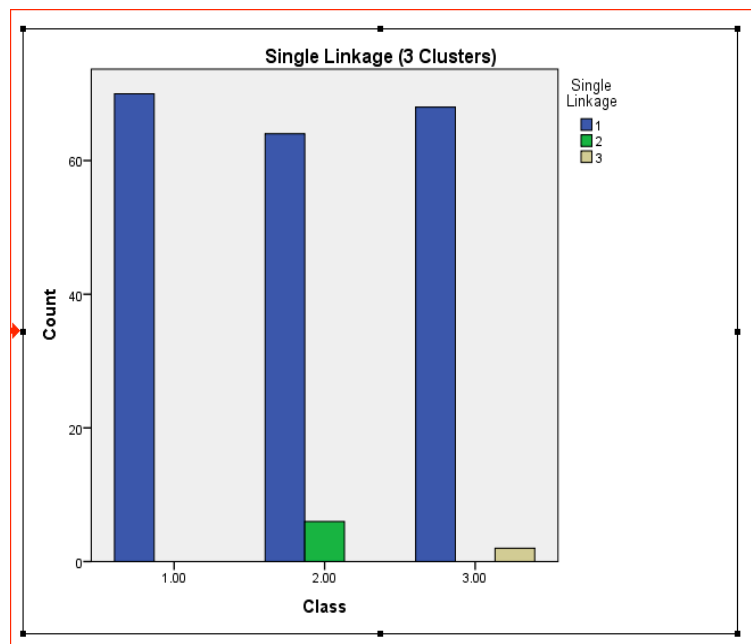
Question3b - 1: Single Linkage algorithms and Reports



Class * Single Linkage Crosstabulation

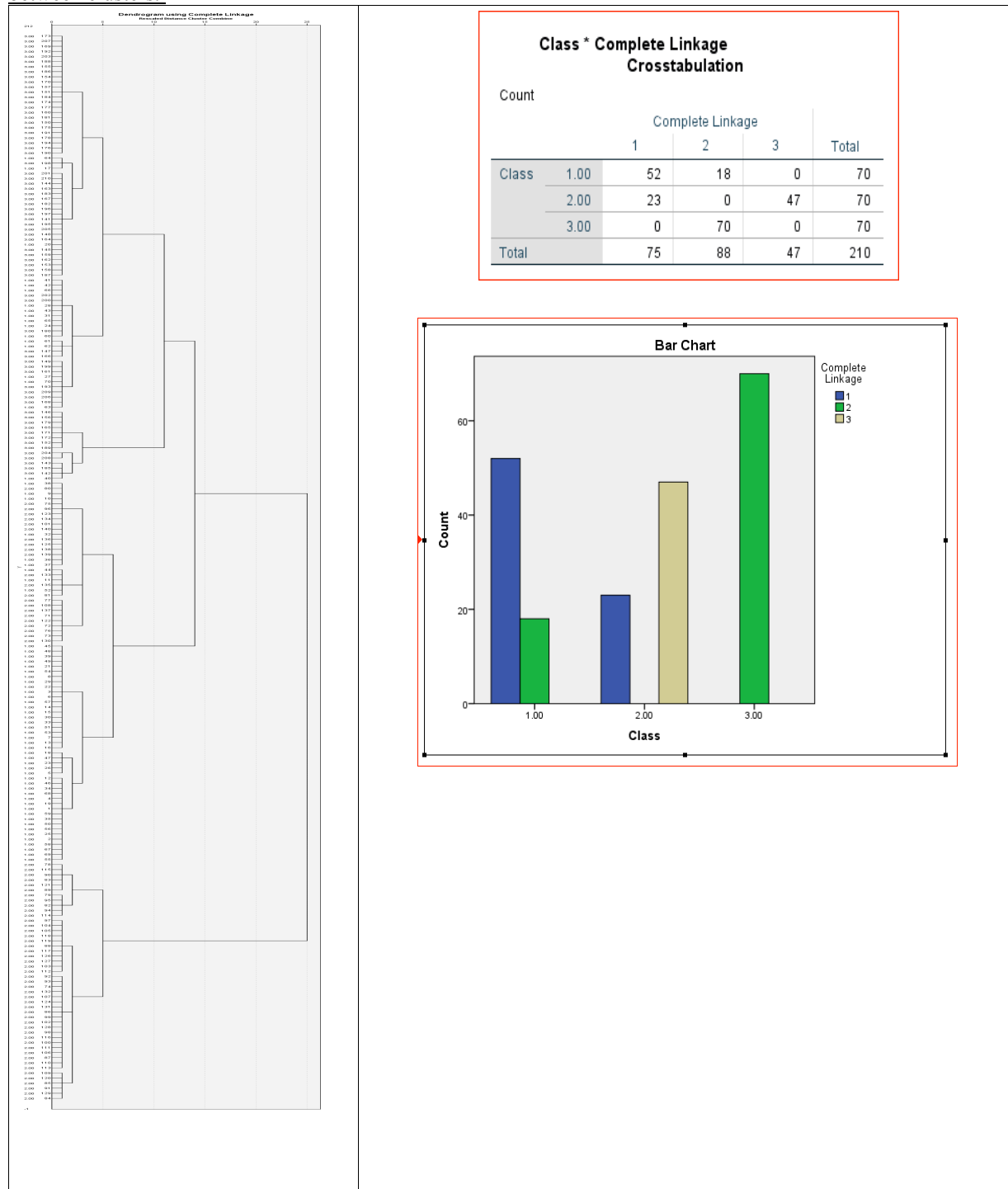
Count

Class	Single Linkage			Total
	1	2	3	
1.00	70	0	0	70
2.00	64	6	0	70
3.00	68	0	2	70
Total	202	6	2	210



Question3b - 2: Complete Linkage algorithms and Reports

Based on the the graph and tables, we can observe that the Complete Linkage Method's result is a lot more accurate than the Single Linkage Method. The Single Link 's result basically only came up 1 cluster(Blue color bar) , but it is supposed to have 3 clusters. On the contrary, the Complete Linkage came up 3 clusters and the predicted values is quite distinguishable between clusters.



Question 3c: Summary

Clustering analysis is the method we have done on the question 3. Clustering analysis is grouping a set of objects together based on their characteristic / similarity. Generally speaking, the objects within a group (cluster) would have the most similar characteristic comparing the objects outside group. The way to group the objects together by using different distance functions (e.g. Euclidean) to calculate the distance between the reference point and the object and to partition objects into a group.

Two major clustering methods which are K-means and Hierarchical analysis were carried out above. K-means analysis is designed to partition the objects/observations into groups (clusters) with the closest cluster centroid (i.e. mean). K-means runs n times iterations to measure the distance of the objects within group and reassign the objects to their nearest cluster centroid. The iterations process normally ends when the clusters reaches the best performance (the distance between objects and centroid within group are shortest). Certainly, there are parameters (e.g. k is the number of clusters) that users have to pre-set before the algorithm start the iteration process. The clusters results could be illustrated by scatter plots, cross-tab matrix and Error graph as shown above. In this particular exercise, the Error graph is used to decide the performance of the clustering analysis. The graph shows the trend of errors changes by increasing the number of clusters. Technically, the more clusters assigned is supposed to lead to better result (less error). So that all we need to do is to observe the graph and find the k (number of cluster) section that starts showing less error improvements. Then that k-means cluster probably the best result of overall analysis.

Hierarchical clustering analysis uses another way to group the similar objects to clusters. Certainly, pre-set parameters (e.g. number of clusters) are necessary prior to the analysis. In this exercise, we have used single linkage and complete linkage algorithms. In the beginning the of the process, the objects are in their clusters of their own, and then the clusters merges by measuring the distance. The clusters combine based on the shortest distance. However, there are different way to define the shortest distance for these 2 methods. The shortest length of single linkage is the shortest distance between 2 members from the clusters; The shortest length of complete linkage is the farthest distance between 2 members from the clusters. With this method, we are not able to check the accuracy. We could just check the dendrogram based on our desired clusters level and class distributions to decide if we would keep the result. For example, if we would like to reach 3 clusters, the result however came up only 2 clusters cases, then we might want to reset our parameters to give another try.

During the K-means clustering analysis, I experienced a very important step to lower the error on the given k (number of clusters) values. Since we are calculated the distance between objects / clusters. So that some skewed numerical attributes values (Too large or small) might influence the distance measures. So that we probably need to normalize the data prior to step in analysis. This would lead us to get the better result. Also, we have to convert the category variables to numerical in order to calculating the distance for clustering.