

# Project 2

## Introduction

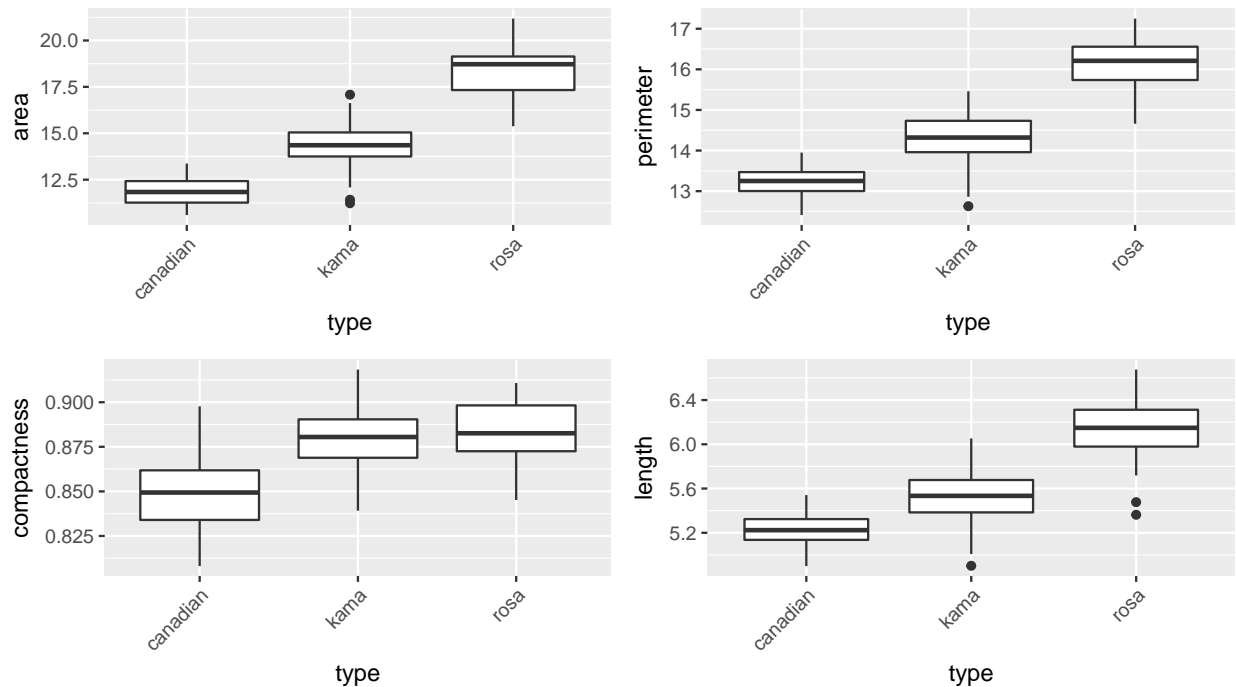
This is Project 2 for STAT 557 2018 Spring by Meridith Bartley and Fei Jiang. The aim of this project is to practice the k-means algorithm and the k-nearest neighbor algorithm. In this project we applied both algorithms to seed data in order to classify/cluster by seed type.

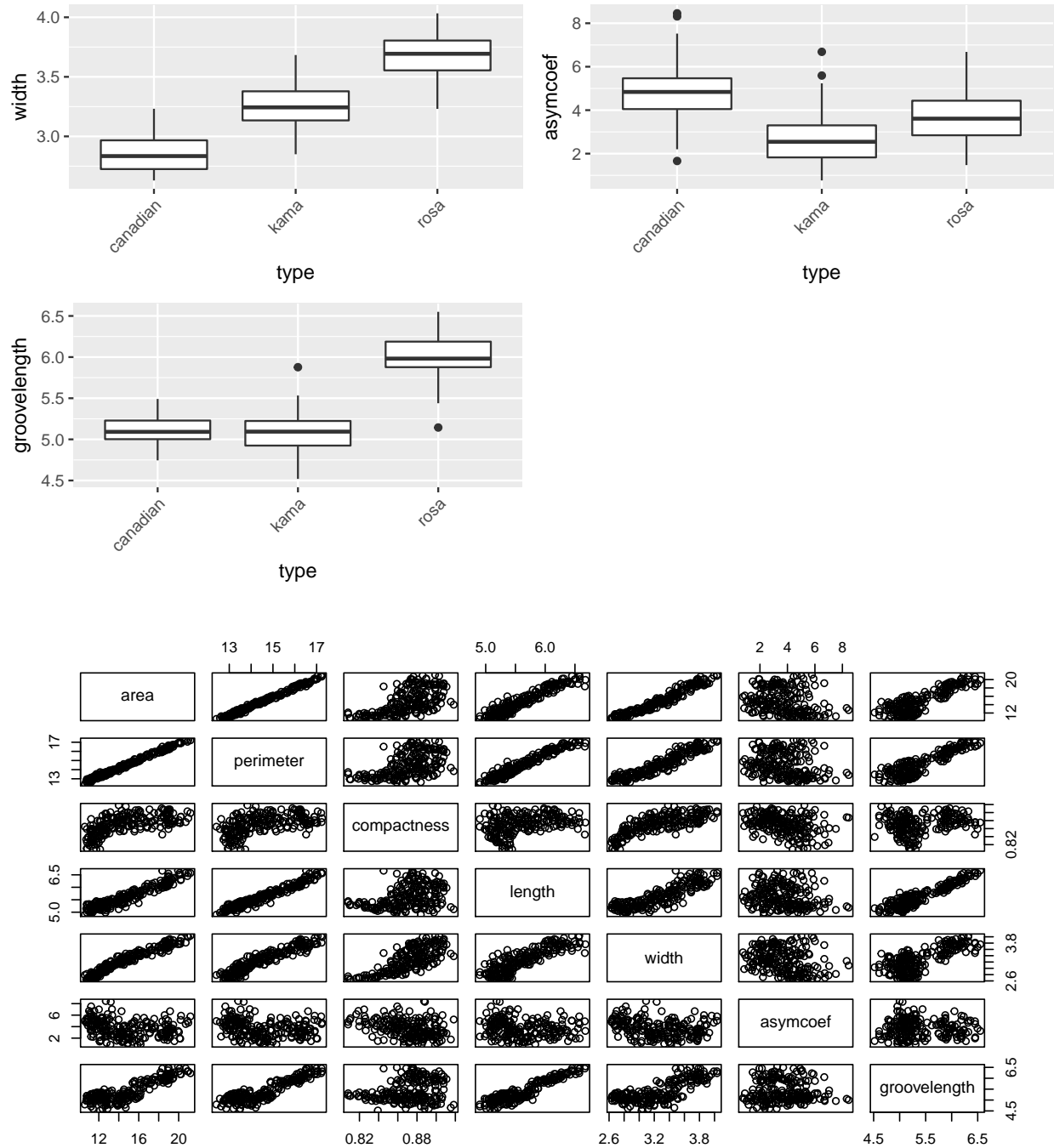
## Description of Data

The examined group comprised kernels belonging to three different varieties of wheat: Kama, Rosa and Canadian, 70 elements each, randomly selected for the experiment. High quality visualization of the internal kernel structure was detected using a soft X-ray technique. It is non-destructive and considerably cheaper than other more sophisticated imaging techniques like scanning microscopy or laser technology. The images were recorded on 13x18 cm X-ray KODAK plates. Studies were conducted using combine harvested wheat grain originating from experimental fields, explored at the Institute of Agrophysics of the Polish Academy of Sciences in Lublin.

Boxplots for each attribute used as explanatory variables in the subsequent classification models are included below. This EDA allows for early indication of which variables may possibly be omitted during dimension reduction. That is, what properties do not differ significantly between seed types.

## Exploratory Data Analysis





## Principle Component Analysis

In order to test whether dimension reduction will improve predictions we also conducted Principle Component Analysis on the original dataset to get a new dataset with fewer dimensions. According to our PCA results, the first two component in total can explain about 99.302% of variance of the original database. The coefficients of the relevent componets are listed in the table below. Therefore, we took the first two components and the seed type values to build a new dataset with less dimensions.

	PC1	PC2	PC3	PC4	PC5	PC6	PC7
Standard deviation	3.28532	1.459265	0.2713485	0.1135231	0.0524235	0.0396289	0.0054457
Proportion of Variance	0.82939	0.163630	0.0056600	0.0009900	0.0002100	0.0001200	0.0000000
Cumulative Proportion	0.82939	0.993020	0.9986800	0.9996700	0.9998800	1.0000000	1.0000000

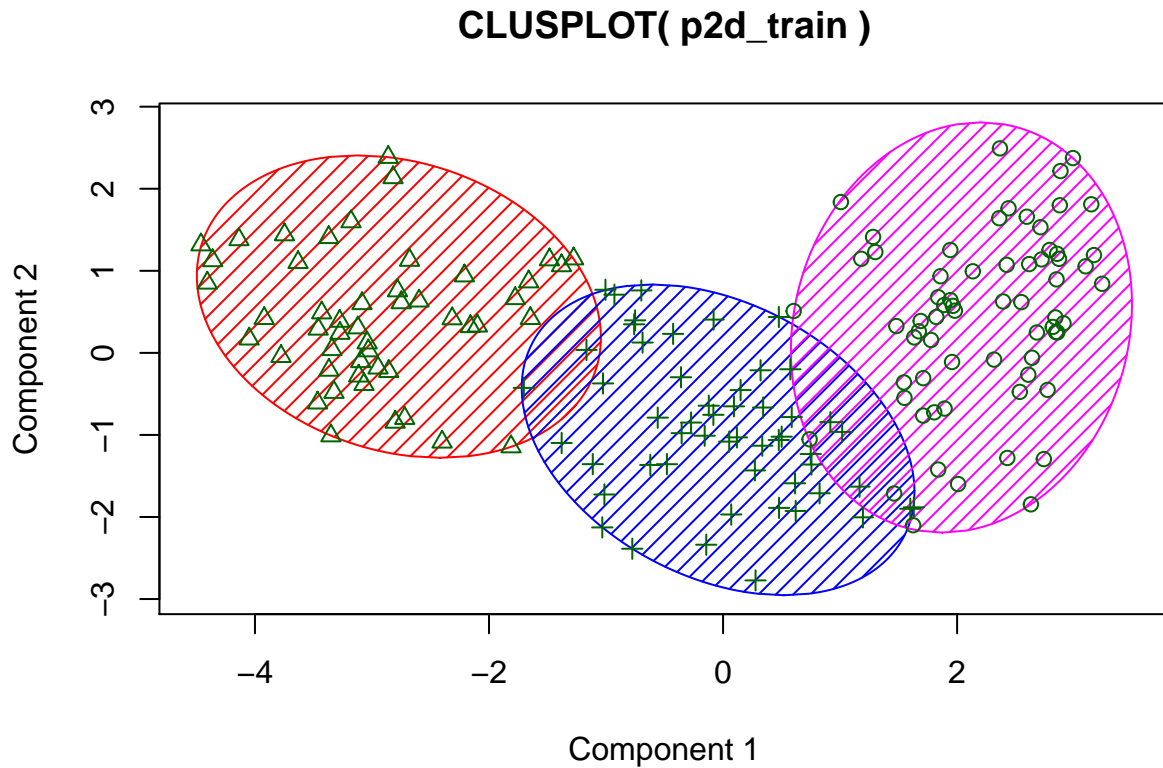
	PC1	PC2
area	0.8842285	0.1008058
perimeter	0.3954054	0.0564896
compactness	0.0043113	-0.0028947
length	0.1285445	0.0306217
width	0.1110591	0.0023723
asymcoef	-0.1276156	0.9894105
groovelength	0.1289665	0.0822334

## Analysis

In the following analysis with two methods (k means and k-nearest neighbor algorithms - both supervised and unsupervised) and two datasets (original and dimension-reduced), we randomly selected 80% of the entire data as training data and the rest 20% as test data.

### K-Means Algorithm - Unsupervised clustering of Original Dataset

When conducting the k-means algorithm with on the original dataset and found that the overall prediction accuracy of our model in testing data is about 90%. In addition, we applied the k-means algorithm to a dimension reduced data set using the first two principal components and found that with the testing data there was about 90% accuracy. In the following plot we can see the cluster plot that uses PCA to draw the data using the first two principal components to explain the data.

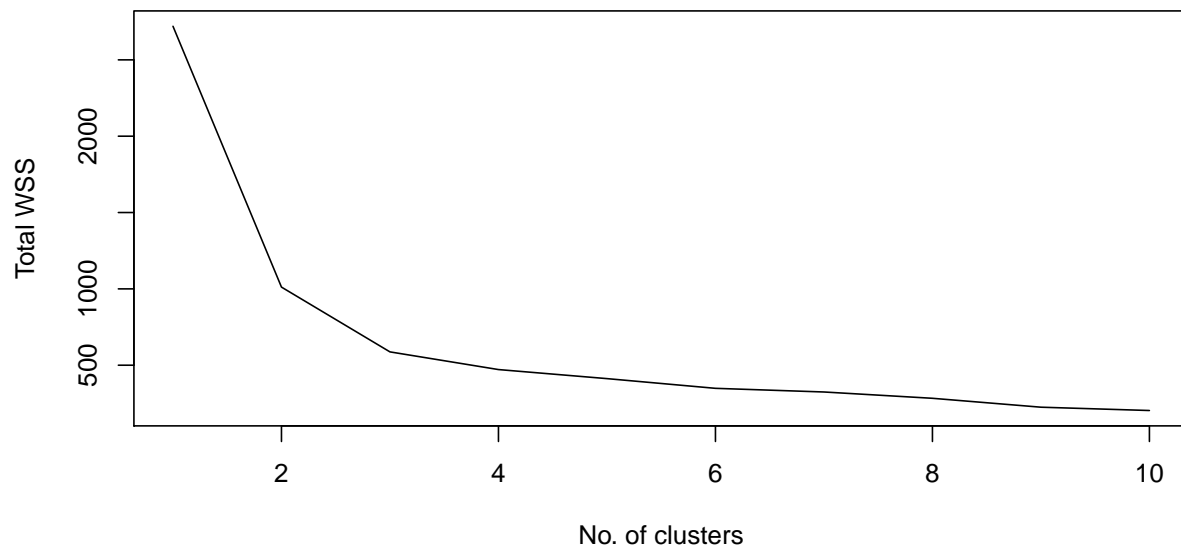


These two components explain 89.24 % of the point variability.

While we have reported the outcomes for  $k = 3$ , we also explored the possibilities of other size selections for  $k$  (see below table for comparison of accuracies). We also looked at total Within Groups Sums of Squares as a measure of how well our clusters fit the data. This is a measure of the distance the vectors in each cluster are from their respected centroid. The goal is to minimize this value but no further than when the rate of improvement drops off. The scree plot below of these WSS values do indicate that is an appropriate number of clusters to choose. Indeed we do know from the data available that there are three seed types included.

K	% Accuracy
2	98
3	90
4	90
5	83
6	93
7	57
8	88
9	95
10	40

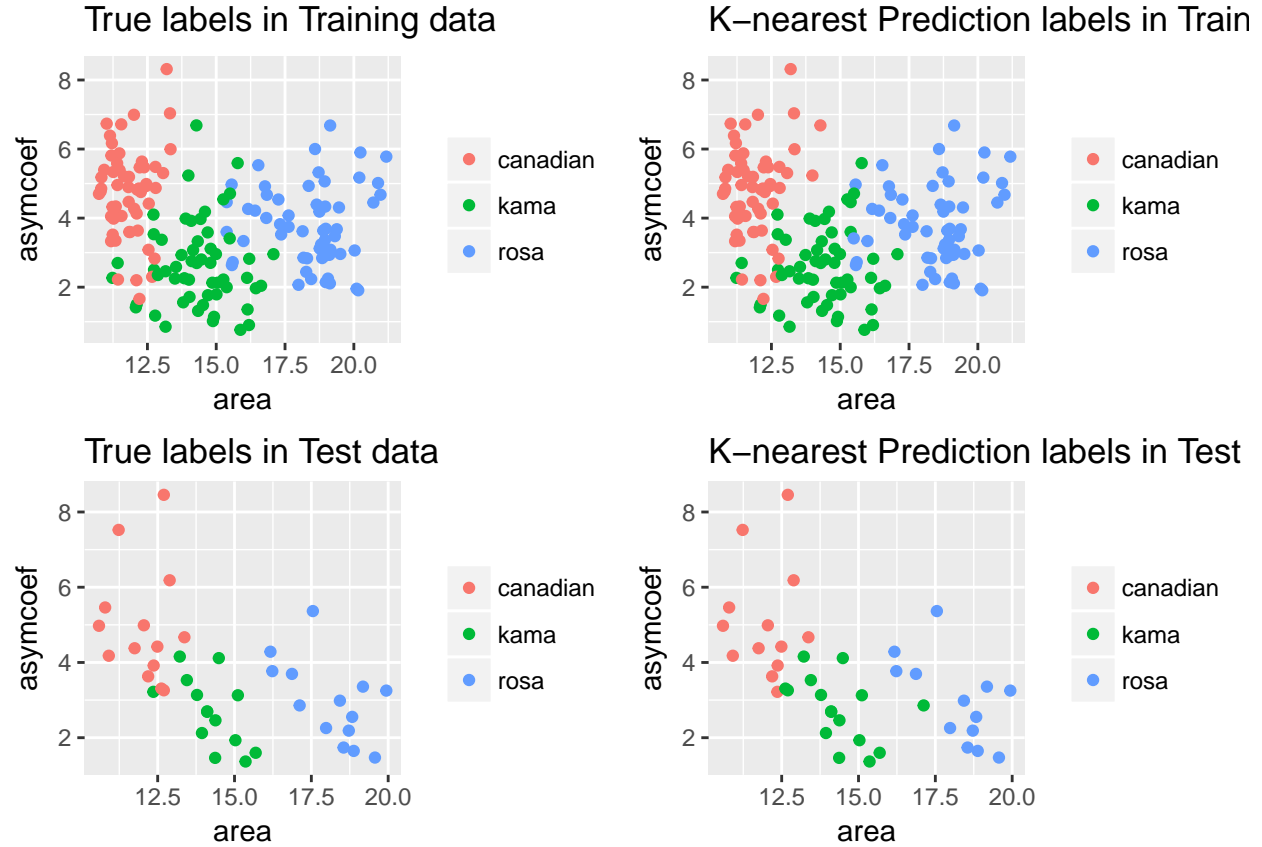
### Scree Plot



### K-Nearest Neighbor - Supervised clustering in Original Dataset

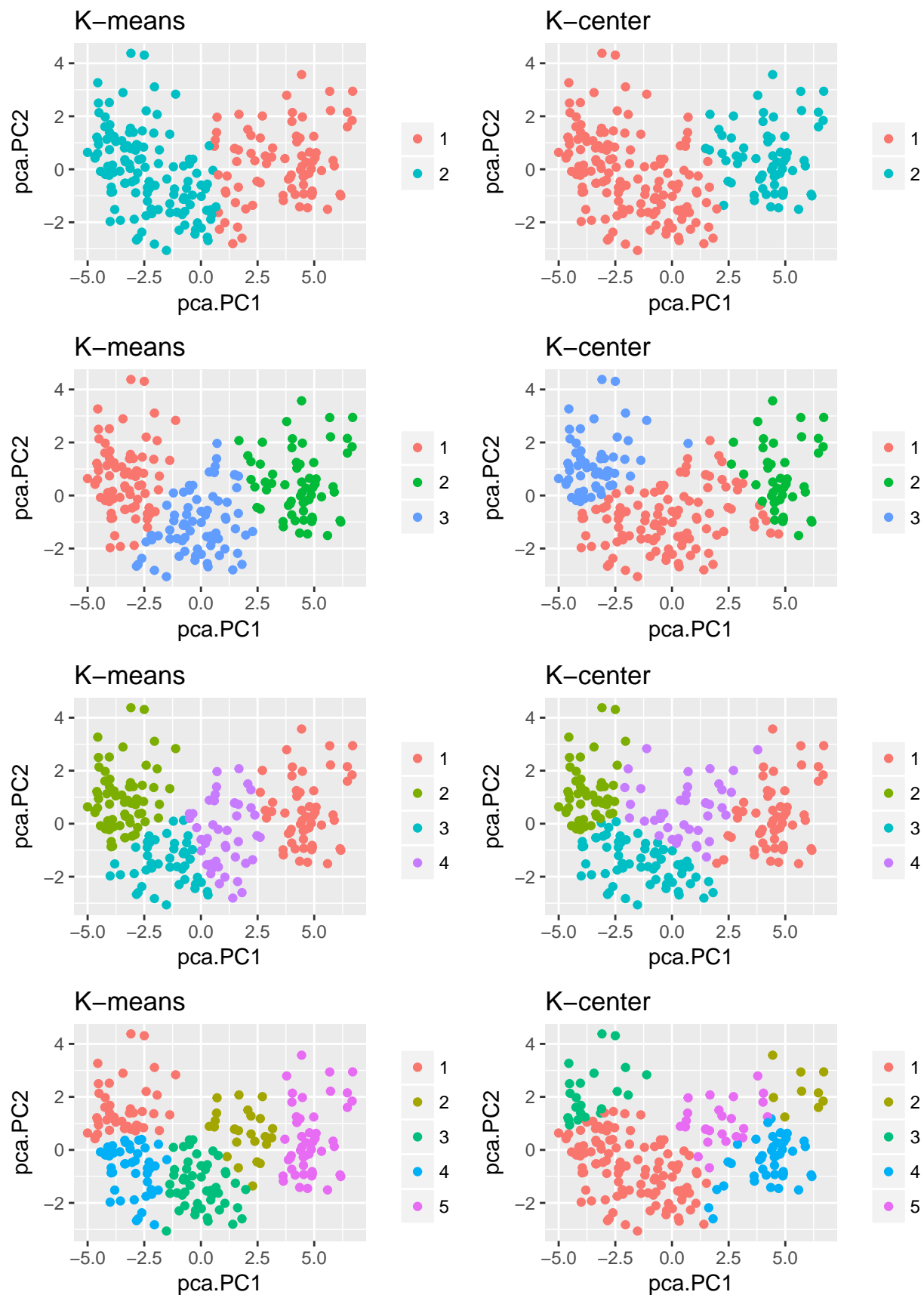
In this section, supervised K-nearest clustering is applied in the original dataset. The true and predicted labels in training and test data are shown in the figures below. In those figures, only first two predictors were shown. In general, in the training data, we obtained the accuracy rate of 92.9% and in the testing data, the accuracy rate is about 90.5%. We think we reached a satisfying accuracy rate.

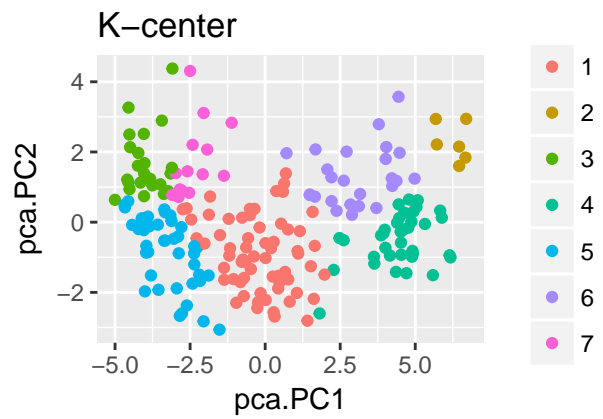
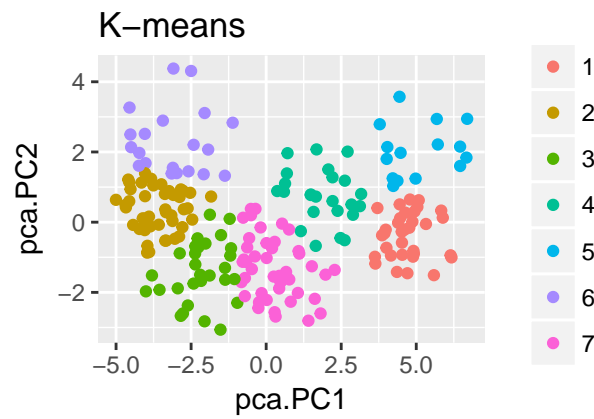
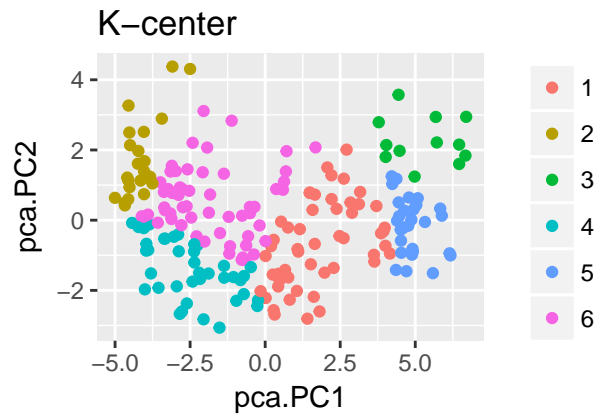
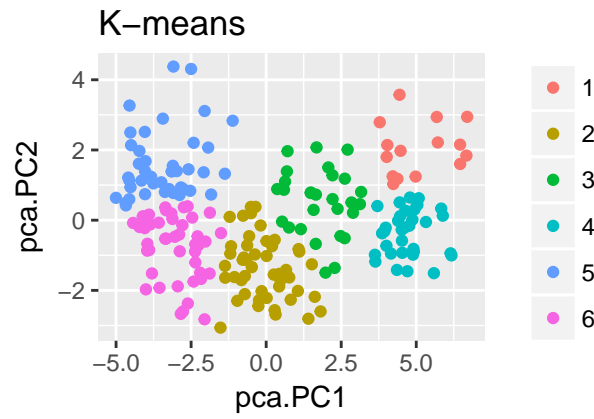
In the following plots we explore the true labels compared to those predicted by K-nearest neighbor clustering. We do this for both the original training and testing datasets. Not that we have plotted the data by area and the asymmetrical coefficient. It's clear that this method of clusing the data is performing well.



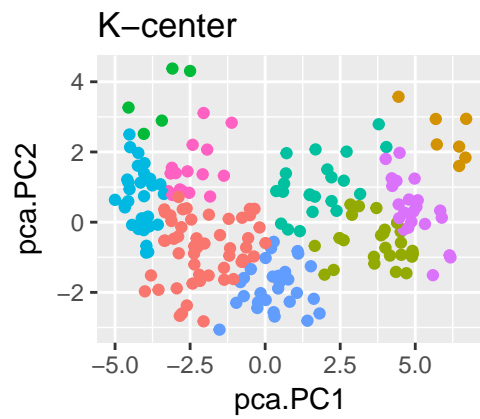
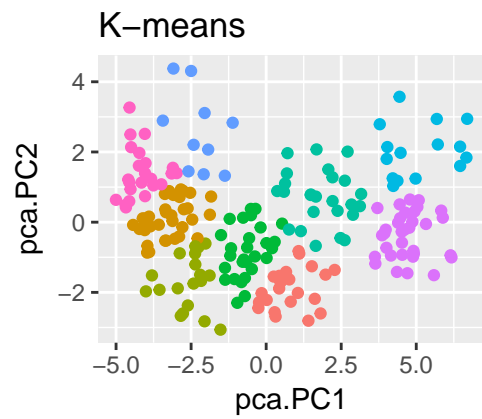
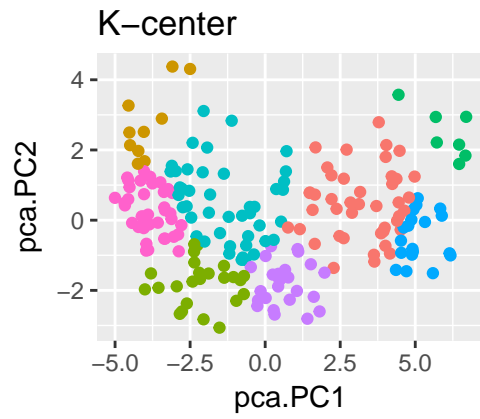
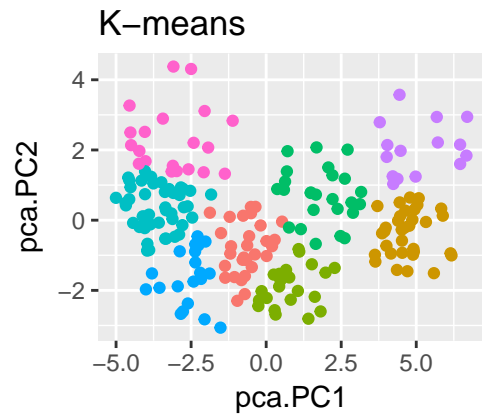
## K-Means and K-Centers Algorithm - Unsupervised clustering in Reduced Dataset (First Two Principle Components)

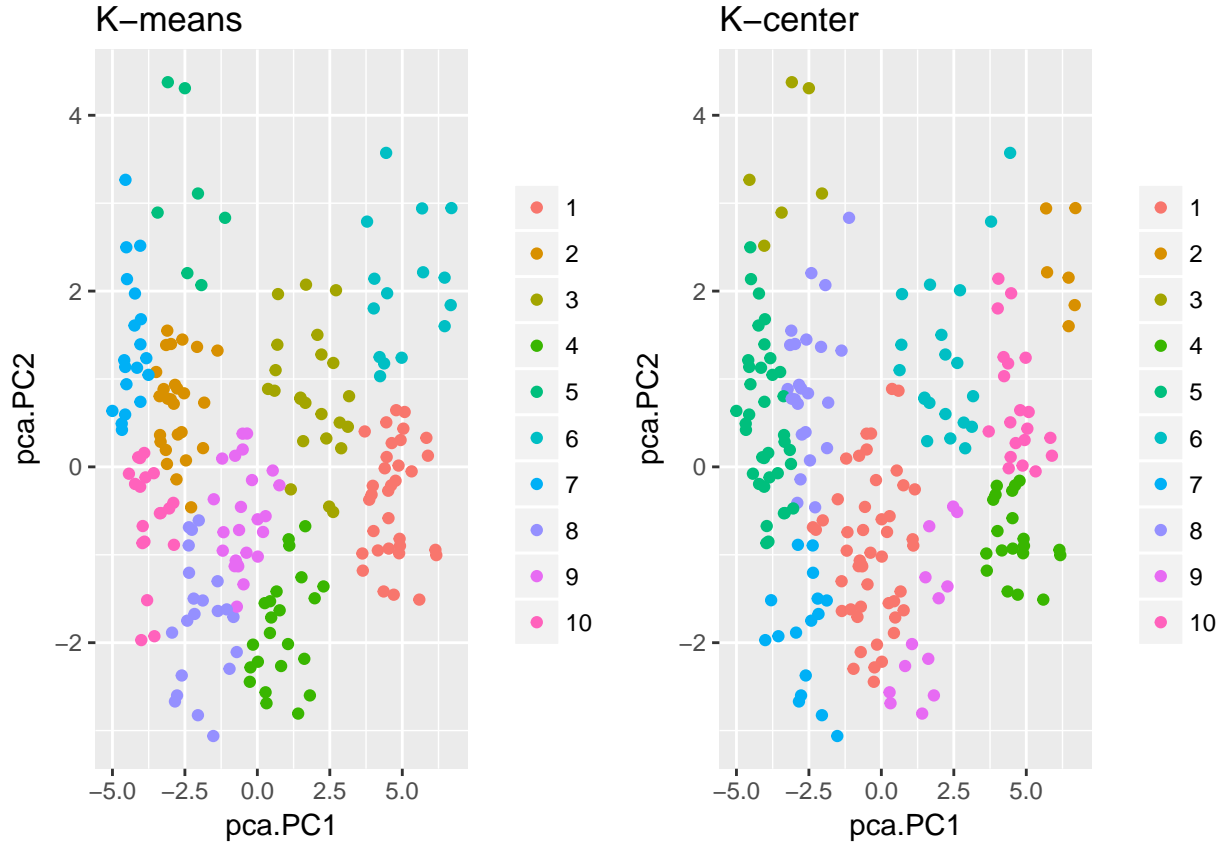
In this section, we applied unsupervised K-Means and K-Centers clustering algorithms to the reduced dataset (first two principle components). We tried 9 different numbers of clusters: 2 to 10 and plotted the scatter plot for each cluster number and for each algorithm. As the below figures show, there are differences between K-Center and K-Means clustering results. That is because k-means focuses on average distance while k-center focuses on worst scenario.











## Conclusions

To sum up, we finished the following analysis in this project.

- The K-means clustering method on both the original and dimension reduced training and testing data, both showing similar accuracies of about 90% for the testing data.
- Confirmed a cluster size of by examining the Within Sums of Squares values.
- The supervised K-nearest clustering method achieved over 90% accuracy in both training and test data, which is satisfying.
- Applying different k values, we explored the contrasting difference between k-means and k-center clustering methods.

## Contributions

The different tasks required to complete this project were equally divided between Meridith and Fei. K-means and cross-validation analyses were completed by Meridith while Fei was responsible for K-nearest and K-center comparison. Both members of this group contributed to this report.