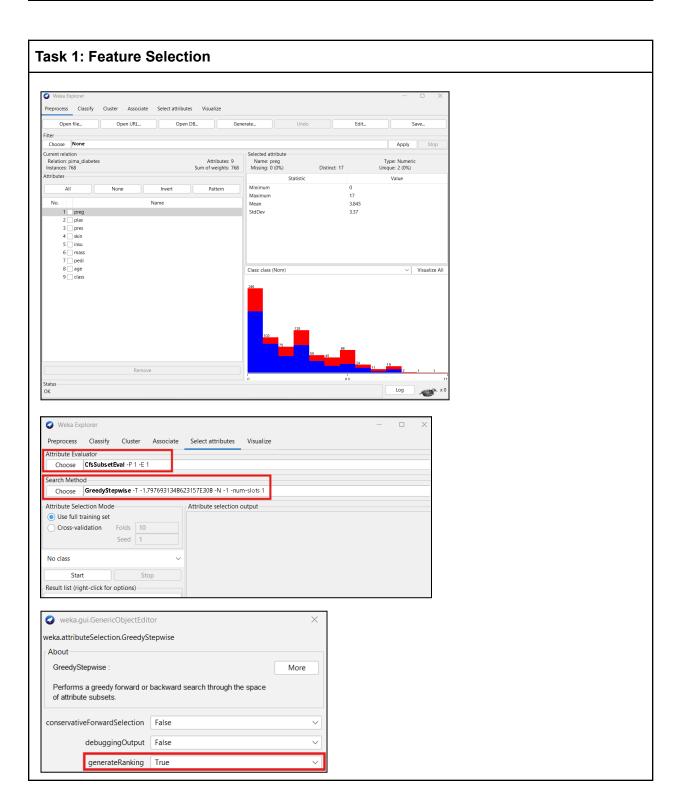
Ardientes, Mary Faith P.	Mr. Kho
IT31S1	09 - 24 - 25



Attribute Selection Mod Use full training set			
	Folds	10	
	Seed	1	

```
Attribute selection output
 === Run information ===
 Evaluator:
             weka.attributeSelection.CfsSubsetEval -P 1 -E 1
              weka.attributeSelection.GreedyStepwise -R -T -1.7976931348623157E308 -N -1 -num-slots 1
 Relation:
           pima_diabetes
              768
 Instances:
Attributes: 9
              preq
              plas
              pres
              skin
              insu
              mass
              pedi
              age
              class
Evaluation mode: evaluate on all training data
 === Attribute Selection on all input data ===
 Search Method:
        Greedy Stepwise (forwards).
        Start set: no attributes
        Ranking is the order that attributes were added, starting
        with no attributes. The merit scores in the left column
         are the goodness of the subset after the adding the
        corresponding attribute in the right column to the subset.
```

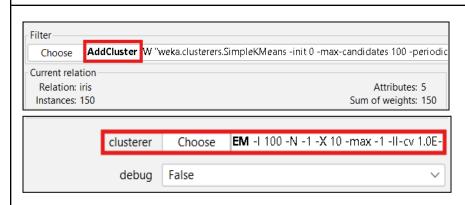
```
Attribute Subset Evaluator (supervised, Class (nominal): 9 class):
       CFS Subset Evaluator
       Including locally predictive attributes
Ranked attributes:
 0.133 2 plas
 0.151 6 mass
 0.164 8 age
 0.161 5 insu
 0.157
       1 preg
 0.153
       7 pedi
 0.147 4 skin
 0.141
        3 pres
Selected attributes: 2,6,8,5,1,7,4,3 : 8
```

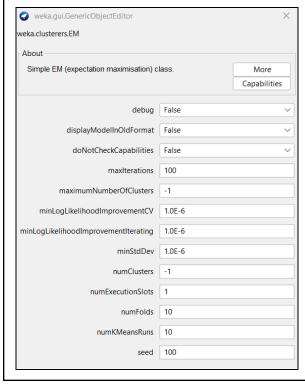
Reflection:

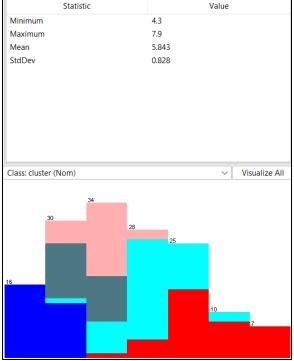
- i. Which features were selected?
 - Following the given procedure, the selected features are CfsSubsetEval with GreedyStepwise; these methods help with predicting the outcome while preventing each other from overlapping and redundancy among them.
- ii. What are the top 3 attributes detected by the CfsSubsetEval algorithm based on their ranked values?
 - Reviewing the output given, the rankings are listed in descending order of importance. The top 3 attributes, based on their merit scores, are
 - Age 0.164
 - Insulin 0.161
 - Mass 0.151

These mean the top 3 attributes that have the highest scores are the most useful for prediction.

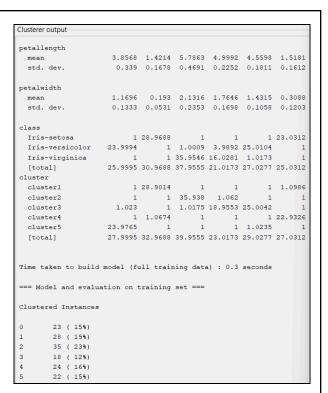
Task 2: Instance Selection

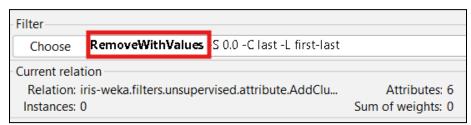


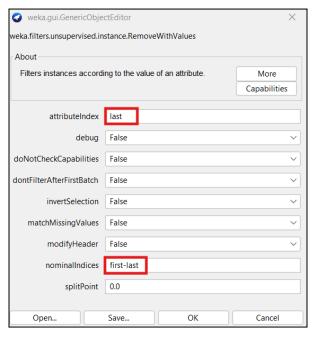




```
weka.clusterers.EM -I 100 -N -1 -X 10 -max -1 -11-cv 1.
Scheme:
             iris-weka.filters.unsupervised.attribute.AddCluster-Wwe
Relation:
Instances:
             150
Attributes:
             sepallength
             sepalwidth
             petallength
             petalwidth
             class
Test mode:
           evaluate on training data
=== Clustering model (full training set) ===
Number of clusters selected by cross validation: 6
Number of iterations performed: 9
                   Cluster
                                      2 3 4
                    (0.15) (0.19) (0.23) (0.12) (0.16) (0.15)
sepallength
                     5.5393 4.7784 6.8577 5.9993 6.2842 5.2949
 std. dev.
                    0.3063 0.2367 0.5168 0.3721 0.3995 0.2365
sepalwidth
                   2.5868 3.182 3.0914 2.7401 2.9247 3.7176
0.2524 0.2566 0.2835 0.2338 0.2713 0.2809
 std. dev.
```





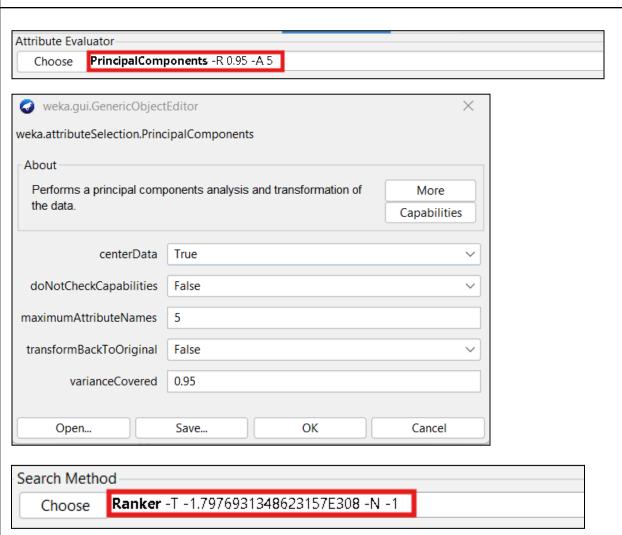




Reflection:

- i. What is the purpose of the newly created attribute "Cluster"?
 - Navigating the filter to choose the cluster is a feature added to the datasat to assign each instance to a cluster based on the patterns shown in the data using the select algorithm (EM). It helps to label instances with a nominal value, to organize them into which cluster they belong to. It overall filters noisy data and understands how the groups of data behave.
- ii. After completing all the procedures, did you encounter the label "Number of clusters selected by cross-validation:"? What does this value represent?
 - Yes, after completing the following procedure, it displayed the number of clusters selected by cross-validation: 6. This represents the number of clusters that EM determined using cross-validation, which evaluates different cluster counts and selects the one that maximizes model likelihood while avoiding overfitting. The EM set follows starts with a default setting og "-1" for the number of clusters, which tells it to let the data decide how many clusters best fit the distribution.

Task 3: Feature Transformation



```
Attribute selection output
=== Run information ===
           weka.attributeSelection.PrincipalComponents -C -R 0.95 -A 5
           weka.attributeSelection.Ranker -T -1.7976931348623157E308 -N -1
Search:
Relation:
          150
Instances:
            sepallength
           sepalwidth
           petallength
           petalwidth
            class
Evaluation mode: evaluate on all training data
=== Attribute Selection on all input data ===
Search Method:
      Attribute ranking.
Attribute Evaluator (unsupervised):
      Principal Components Attribute Transformer
Covariance matrix
 0.69 -0.04 1.27 0.52 -0.28 0.03 0.25
-0.04 0.19 -0.32 -0.12 0.12 -0.1 -0.03
 1.27 -0.32 3.11 1.3 -0.77 0.17 0.6
 0.52 -0.12 1.3 0.58 -0.32 0.04 0.28
 -0.28 0.12 -0.77 -0.32 0.22 -0.11 -0.11
 0.03 -0.1 0.17 0.04 -0.11 0.22 -0.11
 0.25 -0.03 0.6 0.28 -0.11 -0.11 0.22
```

```
eigenvalue proportion cumulative
                                             0.86605
0.07633
0.03863
                                                                                                 0.86605 0.827petallength+0.347petalwidth+0.346sepallength-0.203class=Iris-setosa+0.162class=Iris-virginica...
0.94238 0.687class=Iris-versicolor-0.445sepalwidth-0.416class=Iris-virginica-0.274sepallength-0.271class=Iris-setosa...
0.98101 0.677sepallength+0.469sepalwidth-0.397class=Iris-virginica+0.348class=Iris-versicolor-0.183petalwidth...
     4.53848
    0.20244
 Eigenvectors
  V1 V2 V3
0.3464 -0.274 0.6767 sepallength
  -0.0813 -0.445 0.4693 sepalwidth
   0.8268 0.0763 -0.0879 petallength
   0.3469 -0.0582 -0.1829 petalwidth
   -0.2034 -0.2712 0.0489 class=Iris-setosa
   0.041 0.6867 0.3477 class=Iris-versicolor
  0.1624 -0.4155 -0.3966 class=Iris-virginica
   0.1339-10.827 petallength + 0.347 petalwidth + 0.346 sepallength - 0.203 class = Iris-setosa + 0.162 class = Iris-virginica... + 0.162 class
   0.0576 2 0.687class=Iris-versicolor-0.445sepalwidth-0.416class=Iris-virginica-0.274sepallength-0.271class=Iris-setosa...
   0.019 3 0.677sepallength+0.469sepalwidth-0.397class=Iris-virginica+0.348class=Iris-versicolor-0.183petalwidth...
 Selected attributes: 1,2,3 : 3
```

Reflection:

- i. How many principal components were created, and what percentage of total variance do they explain?
 - Upon observing the output, three principal outcomes are evident in the proportions, which indicate the total variance in the dataset. These values are 0.86605 (86.605%), 0.07633 (7.633%), 0.03863 (3.863%), which explains about 98% (0.98101) of the data's variation, capturing all important patterns, making the dimensionality reduction effective.