# Homework #3

Assign date: 2015-06-21

**Due date: 2015-07-01, 6pm**

Submission:

1. **Please submit your results in email to the grader:** [**130301039@svuca.edu**](mailto:130301039@svuca.edu) **Chi Zhang**
2. **Please separate the written answers from the python code: you should submit 2 files in your email – cs596-29-hw2\_yourID#.doc & cs596-29-hw2\_yourID#.py**
3. **30 pts per day will be deducted for late submission**

**Problem 1)** K-means clustering (20 pts.)

You are given the following 10 data points of height & weight:

|  |  |  |  |  |  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- | --- | --- | --- | --- | --- |
| ID | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 |
| Height | 66 | 73 | 72 | 70 | 74 | 68 | 65 | 64 | 63 | 67 |
| Weight | 170 | 210 | 165 | 180 | 185 | 155 | 150 | 120 | 125 | 140 |

Manually apply k-means algorithms to get 2 clusters. Please produce the center and grouping step by step, using the following parameters:

1. Initialize with ID=1 and ID=2
2. Assume Euclidean distance

**Solution**

In each iteration (EPOCH), we will calculate the Euclidean distance matrix from the 2 seeds to each node. The distance matrix will be used for comparison and inferring if a node belongs to Cluster 1 (represented by Seed 1) or Cluster 2 (represented by Seed 2), the result is recorded in last line of each matrix. Base on the result, new seeds will be calculated and used for next iteration until the Clusters do not change.

**EPOCH 1**

**Seed 1: [66, 170] ∈ C1**

**Seed 2: [73, 210] ∈ C2**

**Distance between 2 seeds to each element:**

**+---------+------+------+------+------+------+------+------+------+------+-------+**

**| | ID=1 | ID=2 | ID=3 | ID=4 | ID=5 | ID=6 | ID=7 | ID=8 | ID=9 | ID=10 |**

**+=========+======+======+======+======+======+======+======+======+======+=======+**

**| Seed-1 | 0 | 40.6 | 7.8 | 10.8 | 17 | 15.1 | 20.0 | 50.0 | 45.1 | 30.0 |**

**+---------+------+------+------+------+------+------+------+------+------+-------+**

**| Seed-2 | 40.6 | 0 | 45.0 | 30.1 | 25.0 | 55.2 | 60.5 | 90.5 | 85.6 | 70.3 |**

**+---------+------+------+------+------+------+------+------+------+------+-------+**

**| Cluster | C1 | C2 | C1 | C1 | C1 | C1 | C1 | C1 | C1 | C1 |**

**+---------+------+------+------+------+------+------+------+------+------+-------+**

**–> Base on above result of new clusterings, new seeds (mean) are computed: [67.66666666666667, 154.44444444444446], and [73.0, 210.0]**

**EPOCH 2**

**Seed 1: [67.66666666666667, 154.44444444444446] ∈ C1**

**Seed 2: [73.0, 210.0] ∈ C2**

**Distance between 2 seeds to each element:**

**+---------+------+------+------+------+------+------+------+------+------+-------+**

**| | ID=1 | ID=2 | ID=3 | ID=4 | ID=5 | ID=6 | ID=7 | ID=8 | ID=9 | ID=10 |**

**+=========+======+======+======+======+======+======+======+======+======+=======+**

**| Seed-1 | 15.6 | 55.8 | 11.4 | 25.7 | 31.2 | 0.7 | 5.2 | 34.6 | 29.8 | 14.5 |**

**+---------+------+------+------+------+------+------+------+------+------+-------+**

**| Seed-2 | 40.6 | 0 | 45.0 | 30.1 | 25.0 | 55.2 | 60.5 | 90.5 | 85.6 | 70.3 |**

**+---------+------+------+------+------+------+------+------+------+------+-------+**

**| Cluster | C1 | C2 | C1 | C1 | C2 | C1 | C1 | C1 | C1 | C1 |**

**+---------+------+------+------+------+------+------+------+------+------+-------+**

**–> Base on above result of new clusterings, new seeds (mean) are computed: [66.875, 150.625], and [73.5, 197.5]**

**EPOCH 3**

**Seed 1: [66.875, 150.625] ∈ C1**

**Seed 2: [73.5, 197.5] ∈ C2**

**Distance between 2 seeds to each element:**

**+---------+------+------+------+------+------+------+------+------+------+-------+**

**| | ID=1 | ID=2 | ID=3 | ID=4 | ID=5 | ID=6 | ID=7 | ID=8 | ID=9 | ID=10 |**

**+=========+======+======+======+======+======+======+======+======+======+=======+**

**| Seed-1 | 19.4 | 59.7 | 15.3 | 29.5 | 35.1 | 4.5 | 2.0 | 30.8 | 25.9 | 10.6 |**

**+---------+------+------+------+------+------+------+------+------+------+-------+**

**| Seed-2 | 28.5 | 12.5 | 32.5 | 17.9 | 12.5 | 42.9 | 48.2 | 78.1 | 73.3 | 57.9 |**

**+---------+------+------+------+------+------+------+------+------+------+-------+**

**| Cluster | C1 | C2 | C1 | C2 | C2 | C1 | C1 | C1 | C1 | C1 |**

**+---------+------+------+------+------+------+------+------+------+------+-------+**

**–> Base on above result of new clusterings, new seeds (mean) are computed: [66.42857142857143, 146.42857142857142], and [72.33333333333333, 191.66666666666666]**

**EPOCH 4**

**Seed 1: [66.42857142857143, 146.42857142857142] ∈ C1**

**Seed 2: [72.33333333333333, 191.66666666666666] ∈ C2**

**Distance between 2 seeds to each element:**

**+---------+------+------+------+------+------+------+------+------+------+-------+**

**| | ID=1 | ID=2 | ID=3 | ID=4 | ID=5 | ID=6 | ID=7 | ID=8 | ID=9 | ID=10 |**

**+=========+======+======+======+======+======+======+======+======+======+=======+**

**| Seed-1 | 23.6 | 63.9 | 19.4 | 33.8 | 39.3 | 8.7 | 3.9 | 26.5 | 21.7 | 6.5 |**

**+---------+------+------+------+------+------+------+------+------+------+-------+**

**| Seed-2 | 22.6 | 18.4 | 26.7 | 11.9 | 6.9 | 36.9 | 42.3 | 72.2 | 67.3 | 51.9 |**

**+---------+------+------+------+------+------+------+------+------+------+-------+**

**| Cluster | C2 | C2 | C1 | C2 | C2 | C1 | C1 | C1 | C1 | C1 |**

**+---------+------+------+------+------+------+------+------+------+------+-------+**

**–> Base on above result of new clusterings, new seeds (mean) are computed: [66.5, 142.5], and [70.75, 186.25]**

**EPOCH 5**

**Seed 1: [66.5, 142.5] ∈ C1**

**Seed 2: [70.75, 186.25] ∈ C2**

**Distance between 2 seeds to each element:**

**+---------+------+------+------+------+------+------+------+------+------+-------+**

**| | ID=1 | ID=2 | ID=3 | ID=4 | ID=5 | ID=6 | ID=7 | ID=8 | ID=9 | ID=10 |**

**+=========+======+======+======+======+======+======+======+======+======+=======+**

**| Seed-1 | 27.5 | 67.8 | 23.2 | 37.7 | 43.2 | 12.6 | 7.7 | 22.6 | 17.9 | 2.5 |**

**+---------+------+------+------+------+------+------+------+------+------+-------+**

**| Seed-2 | 16.9 | 23.9 | 21.3 | 6.3 | 3.5 | 31.4 | 36.7 | 66.6 | 61.7 | 46.4 |**

**+---------+------+------+------+------+------+------+------+------+------+-------+**

**| Cluster | C2 | C2 | C2 | C2 | C2 | C1 | C1 | C1 | C1 | C1 |**

**+---------+------+------+------+------+------+------+------+------+------+-------+**

**–> Base on above result of new clusterings, new seeds (mean) are computed: [65.4, 138.0], and [71.0, 182.0]**

**EPOCH 6**

**Seed 1: [65.4, 138.0] ∈ C1**

**Seed 2: [71.0, 182.0] ∈ C2**

**Distance between 2 seeds to each element:**

**+---------+------+------+------+------+------+------+------+------+------+-------+**

**| | ID=1 | ID=2 | ID=3 | ID=4 | ID=5 | ID=6 | ID=7 | ID=8 | ID=9 | ID=10 |**

**+=========+======+======+======+======+======+======+======+======+======+=======+**

**| Seed-1 | 32.0 | 72.4 | 27.8 | 42.2 | 47.8 | 17.2 | 12.0 | 18.1 | 13.2 | 2.6 |**

**+---------+------+------+------+------+------+------+------+------+------+-------+**

**| Seed-2 | 13 | 28.1 | 17.0 | 2.2 | 4.2 | 27.2 | 32.6 | 62.4 | 57.6 | 42.2 |**

**+---------+------+------+------+------+------+------+------+------+------+-------+**

**| Cluster | C2 | C2 | C2 | C2 | C2 | C1 | C1 | C1 | C1 | C1 |**

**+---------+------+------+------+------+------+------+------+------+------+-------+**

**–> Base on above result of new clusterings, new seeds (mean) are computed: [65.4, 138.0], and [71.0, 182.0]**

After we **Epoch 6**, the Cluster members do not change anymore, so we know Epoch 5/6 has the final result with 2 clusters:

* **Cluster 1: Node {6, 7, 8, 9, 10}**
* **Cluster 2: Node {1, 2, 3, 4, 5}**

**Problem 2)** Agglomerative clustering (20 pts.)

Using the same 10 data points of height & weight as in Problem-1, apply agglomerative clustering manually. Produce the distance matrix and the resulting dendogram. Use the following assumptions:

1. Euclidean distance
2. Single linkage for cluster dissimilarity

**Solution**

**Agglomerative**: initially every point is a cluster of its own and we merge cluster until we end-up with one unique cluster containing all points.

**Single link:** distance between two clusters is the shortest distance between a pair of elements from the two clusters.

**For each k – we recalculate the distance matrix, and merge 2 nodes into 1 node (representing new cluster).**

**We start with k=10, i.e. each point is a cluster.**

k=10

+------+------+------+------+------+------+------+------+------+------+

| 0 | 40.6 | 7.81 | 10.7 | 17.0 | 15.1 | 20.0 | 50.0 | 45.0 | 30.0 |

| | 0788 | 0249 | 7032 | | 3274 | 2498 | 3998 | 9988 | 1666 |

| | 1008 | 6759 | 9614 | | 5950 | 4394 | 4012 | 9135 | 2039 |

| | 5 | 1 | 3 | | 4 | 5 | 8 | 1 | 6 |

+======+======+======+======+======+======+======+======+======+======+

| 40.6 | 0 | 45.0 | 30.1 | 25.0 | 55.2 | 60.5 | 90.4 | 85.5 | 70.2 |

| 08 | | 11 | 50 | 20 | 27 | 31 | 49 | 86 | 57 |

+------+------+------+------+------+------+------+------+------+------+

| 7.81 | 45.0 | 0 | 15.1 | 20.1 | 10.7 | 16.5 | 45.7 | 41 | 25.4 |

| 0 | 11 | | 33 | 00 | 70 | 53 | 06 | | 95 |

+------+------+------+------+------+------+------+------+------+------+

| 10.7 | 30.1 | 15.1 | 0 | 6.40 | 25.0 | 30.4 | 60.2 | 55.4 | 40.1 |

| 70 | 50 | 33 | | 3 | 80 | 14 | 99 | 44 | 12 |

+------+------+------+------+------+------+------+------+------+------+

| 17 | 25.0 | 20.1 | 6.40 | 0 | 30.5 | 36.1 | 65.7 | 61 | 45.5 |

| | 20 | 00 | 3 | | 94 | 39 | 65 | | 41 |

+------+------+------+------+------+------+------+------+------+------+

| 15.1 | 55.2 | 10.7 | 25.0 | 30.5 | 0 | 5.83 | 35.2 | 30.4 | 15.0 |

| 33 | 27 | 70 | 80 | 94 | | 1 | 28 | 14 | 33 |

+------+------+------+------+------+------+------+------+------+------+

| 20.0 | 60.5 | 16.5 | 30.4 | 36.1 | 5.83 | 0 | 30.0 | 25.0 | 10.1 |

| 25 | 31 | 53 | 14 | 39 | 1 | | 17 | 80 | 98 |

+------+------+------+------+------+------+------+------+------+------+

| 50.0 | 90.4 | 45.7 | 60.2 | 65.7 | 35.2 | 30.0 | 0 | 5.09 | 20.2 |

| 40 | 49 | 06 | 99 | 65 | 28 | 17 | | 9 | 24 |

+------+------+------+------+------+------+------+------+------+------+

| 45.1 | 85.5 | 41 | 55.4 | 61 | 30.4 | 25.0 | 5.09 | 0 | 15.5 |

| 00 | 86 | | 44 | | 14 | 80 | 9 | | 24 |

+------+------+------+------+------+------+------+------+------+------+

| 30.0 | 70.2 | 25.4 | 40.1 | 45.5 | 15.0 | 10.1 | 20.2 | 15.5 | 0 |

| 17 | 57 | 95 | 12 | 41 | 33 | 98 | 24 | 24 | |

+------+------+------+------+------+------+------+------+------+------+

k=9

+-------+-------+-------+-------+-------+-------+-------+-------+-------+

| 0 | 40.60 | 7.810 | 10.77 | 17.0 | 15.13 | 20.02 | 61.72 | 30.01 |

| | 78810 | 24967 | 03296 | | 27459 | 49843 | 69089 | 66620 |

| | 085 | 591 | 143 | | 504 | 945 | 274 | 396 |

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| 40.60 | 0 | 45.01 | 30.15 | 25.02 | 55.22 | 60.53 | 115.6 | 70.25 |

| 8 | | 1 | 0 | 0 | 7 | 1 | 57 | 7 |

+-------+-------+-------+-------+-------+-------+-------+-------+-------+

| 7.810 | 45.01 | 0 | 15.13 | 20.10 | 10.77 | 16.55 | 56.10 | 25.49 |

| | 1 | | 3 | 0 | 0 | 3 | 4 | 5 |

+-------+-------+-------+-------+-------+-------+-------+-------+-------+

| 10.77 | 30.15 | 15.13 | 0 | 6.403 | 25.08 | 30.41 | 75.46 | 40.11 |

| 0 | 0 | 3 | | | 0 | 4 | 2 | 2 |

+-------+-------+-------+-------+-------+-------+-------+-------+-------+

| 17 | 25.02 | 20.10 | 6.403 | 0 | 30.59 | 36.13 | 82.81 | 45.54 |

| | 0 | 0 | | | 4 | 9 | 0 | 1 |

+-------+-------+-------+-------+-------+-------+-------+-------+-------+

| 15.13 | 55.22 | 10.77 | 25.08 | 30.59 | 0 | 5.831 | 42.06 | 15.03 |

| 3 | 7 | 0 | 0 | 4 | | | 1 | 3 |

+-------+-------+-------+-------+-------+-------+-------+-------+-------+

| 20.02 | 60.53 | 16.55 | 30.41 | 36.13 | 5.831 | 0 | 35.03 | 10.19 |

| 5 | 1 | 3 | 4 | 9 | | | 1 | 8 |

+-------+-------+-------+-------+-------+-------+-------+-------+-------+

| 61.72 | 115.6 | 56.10 | 75.46 | 82.81 | 42.06 | 35.03 | 0 | 22.13 |

| 7 | 57 | 4 | 2 | 0 | 1 | 1 | | 2 |

+-------+-------+-------+-------+-------+-------+-------+-------+-------+

| 30.01 | 70.25 | 25.49 | 40.11 | 45.54 | 15.03 | 10.19 | 22.13 | 0 |

| 7 | 7 | 5 | 2 | 1 | 3 | 8 | 2 | |

+-------+-------+-------+-------+-------+-------+-------+-------+-------+

k=8

+--------+--------+--------+--------+--------+--------+--------+--------+

| 0 | 40.607 | 7.8102 | 10.770 | 17.0 | 21.494 | 61.726 | 30.016 |

| | 881008 | 496759 | 329614 | | 836265 | 908927 | 662039 |

| | 5 | 1 | 3 | | | 4 | 6 |

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| 40.608 | 0 | 45.011 | 30.150 | 25.020 | 75.228 | 115.65 | 70.257 |

| | | | | | | 7 | |

+--------+--------+--------+--------+--------+--------+--------+--------+

| 7.810 | 45.011 | 0 | 15.133 | 20.100 | 16.272 | 56.104 | 25.495 |

+--------+--------+--------+--------+--------+--------+--------+--------+

| 10.770 | 30.150 | 15.133 | 0 | 6.403 | 35.052 | 75.462 | 40.112 |

+--------+--------+--------+--------+--------+--------+--------+--------+

| 17 | 25.020 | 20.100 | 6.403 | 0 | 42.545 | 82.810 | 45.541 |

+--------+--------+--------+--------+--------+--------+--------+--------+

| 21.495 | 75.228 | 16.272 | 35.052 | 42.545 | 0 | 54.904 | 14.877 |

+--------+--------+--------+--------+--------+--------+--------+--------+

| 61.727 | 115.65 | 56.104 | 75.462 | 82.810 | 54.904 | 0 | 22.132 |

| | 7 | | | | | | |

+--------+--------+--------+--------+--------+--------+--------+--------+

| 30.017 | 70.257 | 25.495 | 40.112 | 45.541 | 14.877 | 22.132 | 0 |

+--------+--------+--------+--------+--------+--------+--------+--------+

k=7

+--------+-----------+-----------+-----------+-----------+-----------+-----------+

| 0 | 40.607881 | 7.8102496 | 16.379178 | 21.494836 | 61.726908 | 30.016662 |

| | 0085 | 7591 | 3304 | 265 | 9274 | 0396 |

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| 40.608 | 0 | 45.011 | 34.645 | 75.228 | 115.657 | 70.257 |

+--------+-----------+-----------+-----------+-----------+-----------+-----------+

| 7.810 | 45.011 | 0 | 21.354 | 16.272 | 56.104 | 25.495 |

+--------+-----------+-----------+-----------+-----------+-----------+-----------+

| 16.379 | 34.645 | 21.354 | 0 | 54.996 | 115.503 | 54.968 |

+--------+-----------+-----------+-----------+-----------+-----------+-----------+

| 21.495 | 75.228 | 16.272 | 54.996 | 0 | 54.904 | 14.877 |

+--------+-----------+-----------+-----------+-----------+-----------+-----------+

| 61.727 | 115.657 | 56.104 | 115.503 | 54.904 | 0 | 22.132 |

+--------+-----------+-----------+-----------+-----------+-----------+-----------+

| 30.017 | 70.257 | 25.495 | 54.968 | 14.877 | 22.132 | 0 |

+--------+-----------+-----------+-----------+-----------+-----------+-----------+

k=6

+--------+-------------+-------------+-------------+-------------+-------------+

| 0 | 54.47591060 | 24.39472644 | 24.41990187 | 84.46809174 | 34.40442317 |

| | 68 | 68 | 28 | 33 | 97 |

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| 54.476 | 0 | 34.645 | 75.228 | 115.657 | 70.257 |

+--------+-------------+-------------+-------------+-------------+-------------+

| 24.395 | 34.645 | 0 | 54.996 | 115.503 | 54.968 |

+--------+-------------+-------------+-------------+-------------+-------------+

| 24.420 | 75.228 | 54.996 | 0 | 54.904 | 14.877 |

+--------+-------------+-------------+-------------+-------------+-------------+

| 84.468 | 115.657 | 115.503 | 54.904 | 0 | 22.132 |

+--------+-------------+-------------+-------------+-------------+-------------+

| 34.404 | 70.257 | 54.968 | 14.877 | 22.132 | 0 |

+--------+-------------+-------------+-------------+-------------+-------------+

k=5

+--------+---------------+---------------+---------------+---------------+

| 0 | 54.4759106068 | 24.3947264468 | 34.2276795506 | 84.4680917433 |

+========+===============+===============+===============+===============+

| 54.476 | 0 | 34.645 | 87.830 | 115.657 |

+--------+---------------+---------------+---------------+---------------+

| 24.395 | 34.645 | 0 | 71.027 | 115.503 |

+--------+---------------+---------------+---------------+---------------+

| 34.228 | 87.830 | 71.027 | 0 | 51.252 |

+--------+---------------+---------------+---------------+---------------+

| 84.468 | 115.657 | 115.503 | 51.252 | 0 |

+--------+---------------+---------------+---------------+---------------+

k=4

+---------+---------------+---------------+---------------+

| 0 | 48.5938237752 | 64.7269108126 | 125.182321436 |

+=========+===============+===============+===============+

| 48.594 | 0 | 87.830 | 115.657 |

+---------+---------------+---------------+---------------+

| 64.727 | 87.830 | 0 | 51.252 |

+---------+---------------+---------------+---------------+

| 125.182 | 115.657 | 51.252 | 0 |

+---------+---------------+---------------+---------------+

k=3

+---------+---------------+---------------+

| 0 | 82.3284543958 | 142.982492917 |

+=========+===============+===============+

| 82.328 | 0 | 51.252 |

+---------+---------------+---------------+

| 142.982 | 51.252 | 0 |

+---------+---------------+---------------+

k=2

+---------+---------------+

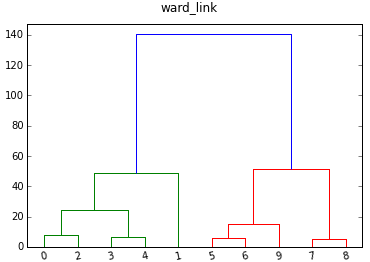
| 0 | 140.324632695 |

+=========+===============+

| 140.325 | 0 |

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**Dendogram**

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**Problem 3)** Model Evaluation (20 pts.)  
We are given the following classification results for a cancer screening test. Please derive the evaluation parameters:

1. Accuracy
2. Specificity
3. Sensitivity
4. Precision
5. Recall

|  |  |  |  |
| --- | --- | --- | --- |
| Actual / Prediction | test = positive | test = negative | Total |
| cancer = yes | **90** | **210** | 300 |
| cancer = no | **140** | **9560** | 9700 |
| Total | 230 | 9770 | 10000 |

**Solution**

|  |  |  |  |
| --- | --- | --- | --- |
| Actual / Prediction | test = positive | test = negative | Total |
| cancer = yes | **TP = 90** | **FN = 210** | 300 |
| cancer = no | **FP = 140** | **TN = 9560** | 9700 |
| Total | 230 | 9770 | 10000 |

TP = 90

FN = 210

FP = 140

TN = 9560

**Accuracy = #correct / N**

**= (**TP + TN) / (TP + FP + TN + FN)

= (90 + 9560) / (90 + 140 + 9560 + 210)

= 0.965

**Specificity = TN / (TN + FP)**

= 9560 / 9700

= 0.986

**Sensitivity = TP / (TP + FN)**

= 90 / 300

= 0.3

**Precision = TP / (TP + FP)**

= 90 / 230

= 0.391

**Recall = Sensitivity**

= 0.3

**F1 = P\*R / 2(P+R)**

= 0.084

**Problem 4)** Python program (40 pts)

Take the sk-learn sample code *cluster\_kmeans.py* discussed in the class that is used for clustering iris dataset. Make the following changes:

1. In addition to using PCA to reduce features from 4 to 2, also evaluate using the original feature pairs (1,2) and (3, 4).
2. For all the 3 clustering settings (original, (1,2), (3, 4)), calculate the clustering quality

CQ = IE / EV as defined in the class:

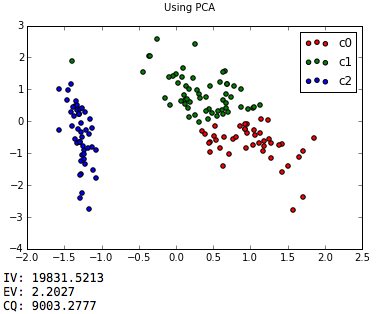
latex-image-1.pdf and latex-image-1.pdf

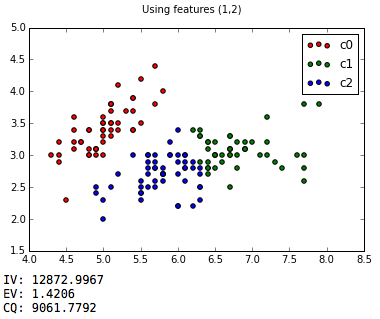
**Output:**

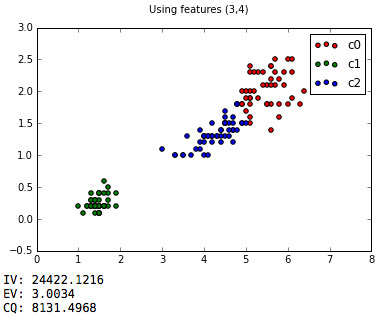
**3 plots of k-means clustering results for PCA, features (1,2), and features (3,4).**

**3 CQ values for PCA, features (1,2), and features (3,4).**

**Solution**

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