

MACHINE LEARNING AND PATTERN RECOGNITION

IMAGE AND VIDEO UNDERSTAND

K-means & Spectral Clustering

Image Segmentation

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Outline

1. K-means
2. Spectral Clustering
3. Elbow method
4. Apply to datasets
5. Image segmentation

K-means clustering

Step 1: Pick K random points as cluster centers

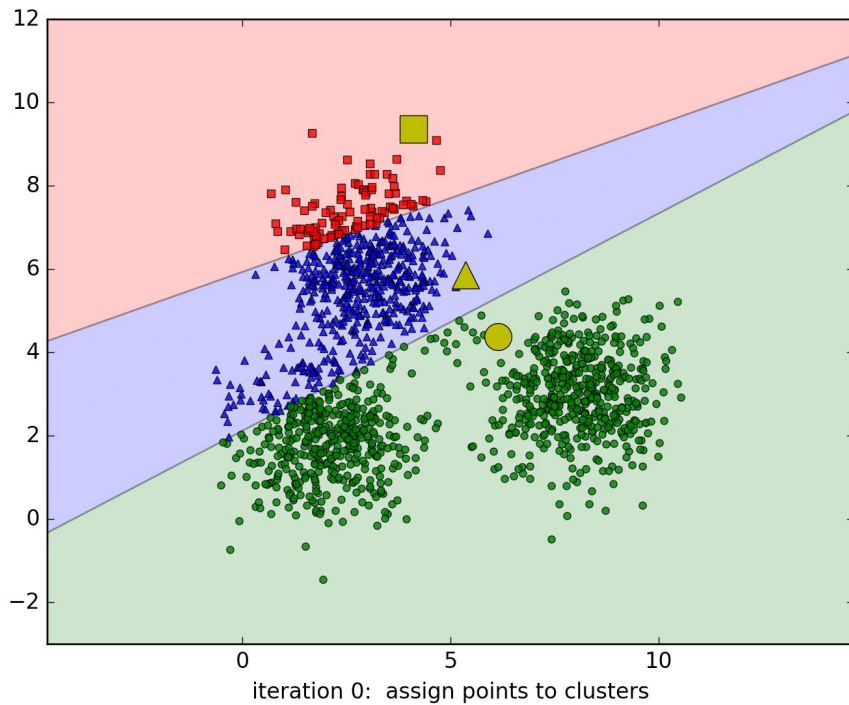
Step 2: Assign data points to the closest centers

Step 3: Change the cluster center to the average of the assigned points

Step 4: Repeat step 2 and 3 until the centers don't change

$$\mathbf{y}_i = \arg \min_{\mathbf{y}_i} \sum_{j=1}^K y_{ij} \|\mathbf{x}_i - \mathbf{m}_j\|_2^2$$

$$\text{subject to: } y_{ij} \in \{0, 1\} \quad \forall j; \quad \sum_{j=1}^K y_{ij} = 1$$



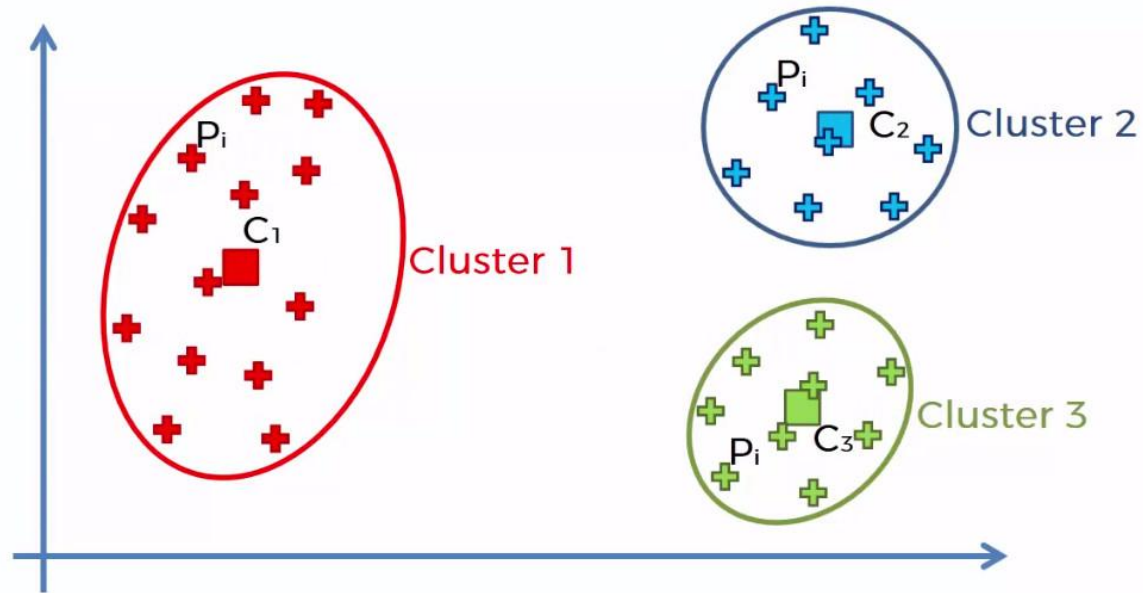
Spectral clustering

Step 1: Construct a similarity graph (KNN graph)

Step 2: Embed the data points in low dimensional space in which the clusters are

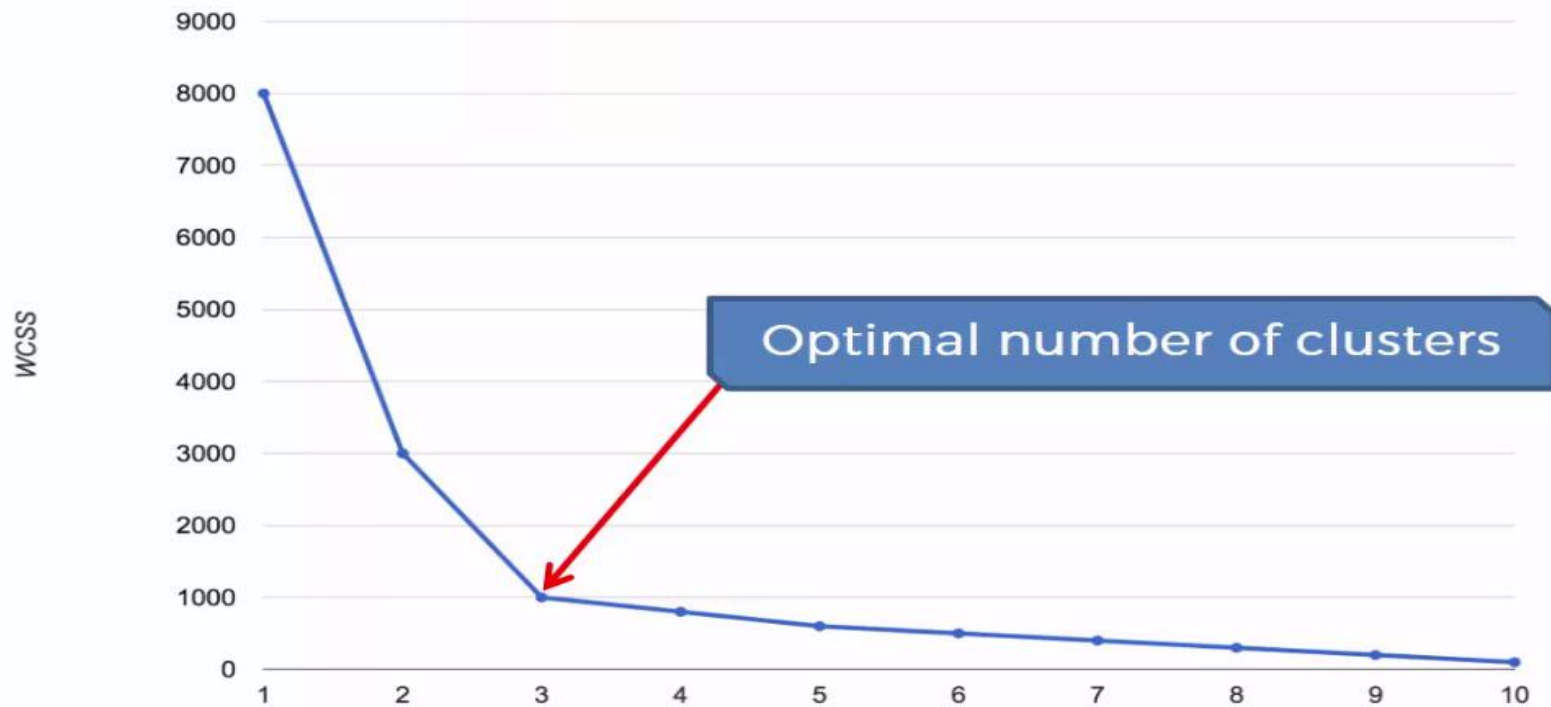
Step 3: Use the lowest eigenvalue in order to choose the eigenvector for clusters

Elbow method



$$WCSS = \sum_{P_i \text{ in Cluster 1}} \text{distance}(P_i, C_1)^2 + \sum_{P_i \text{ in Cluster 2}} \text{distance}(P_i, C_2)^2 + \sum_{P_i \text{ in Cluster 3}} \text{distance}(P_i, C_3)^2$$

The Elbow Method

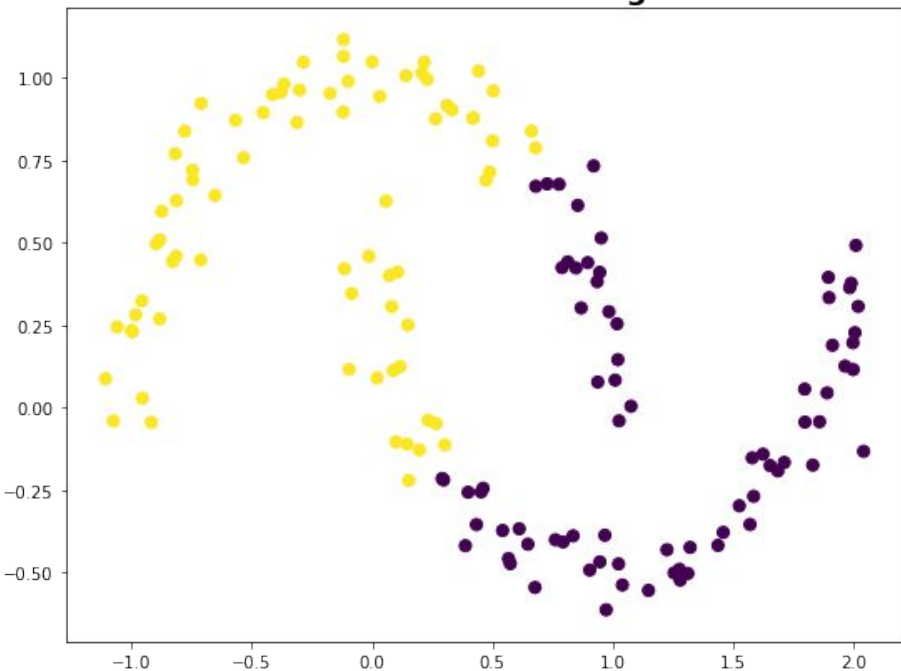


Datasets

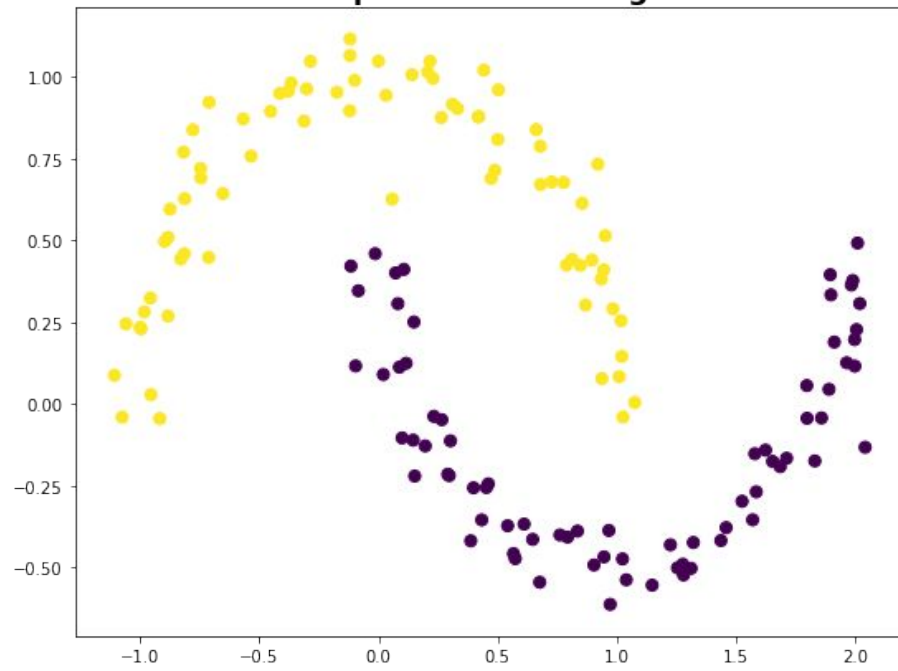
Dataset	Num of instances	Num of attributes	Num of clusters
Moons	150	2	2
Iris	150	4	3
Seeds	210	7	3

Moons dataset

K-means clustering



Spectral clustering



Iris dataset

1. sepal length in cm

2. sepal width in cm

3. petal length in cm

4. petal width in cm

5. Class:

-- Iris Setosa

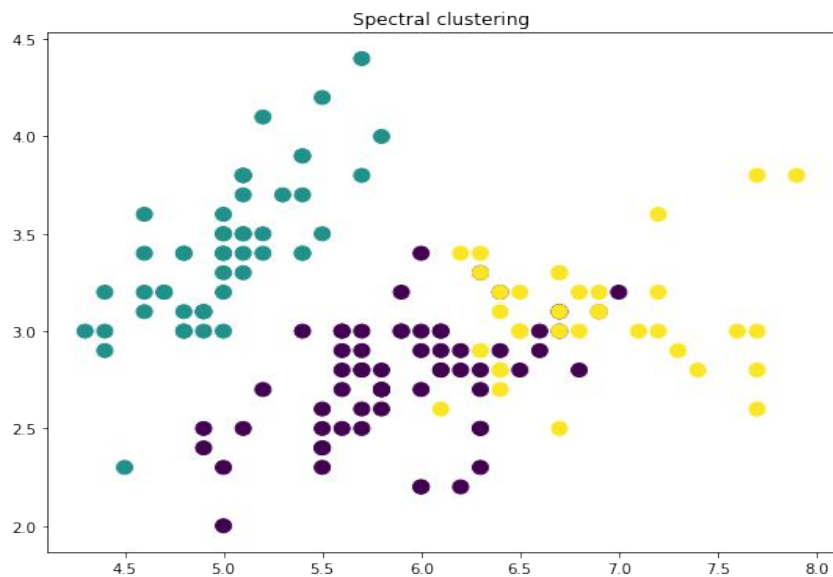
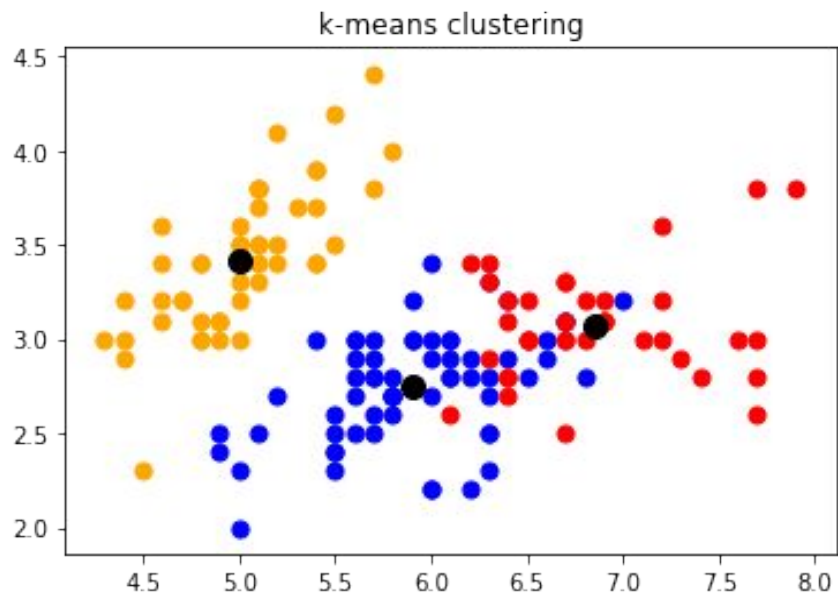
-- Iris Versicolour

-- Iris Virginica

```
dataset.head()
```

	Id	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm	Species
0	1	5.1	3.5	1.4	0.2	Iris-setosa
1	2	4.9	3.0	1.4	0.2	Iris-setosa
2	3	4.7	3.2	1.3	0.2	Iris-setosa
3	4	4.6	3.1	1.5	0.2	Iris-setosa
4	5	5.0	3.6	1.4	0.2	Iris-setosa

Iris dataset



Seeds dataset

- 0. area A,
- 1. perimeter P,
- 2. compactness $C = 4 \cdot \pi \cdot A / P^2$,
- 3. length of kernel,
- 4. width of kernel,
- 5. asymmetry coefficient
- 6. Length of kernel groove.
- 7. Labels

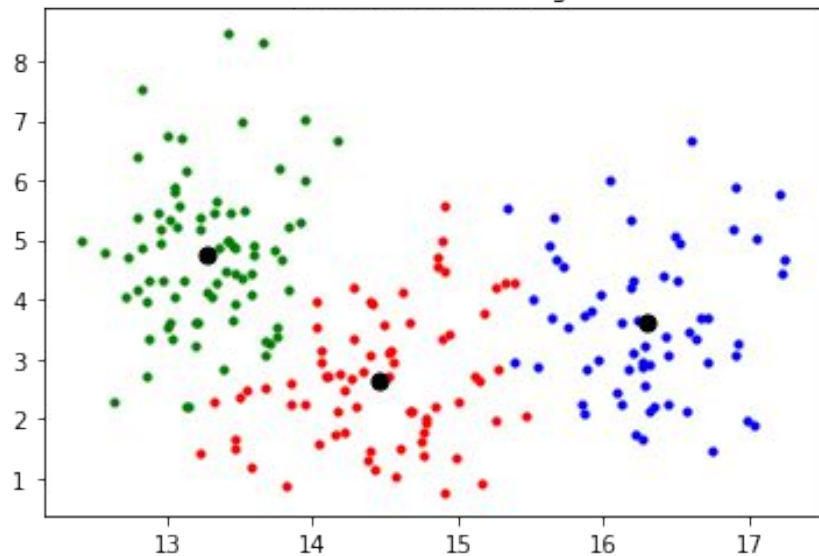
```
data.head()
```

	0	1	2	3	4	5	6	7
0	15.26	14.84	0.8710	5.763	3.312	2.221	5.220	1
1	14.88	14.57	0.8811	5.554	3.333	1.018	4.956	1
2	14.29	14.09	0.9050	5.291	3.337	2.699	4.825	1
3	13.84	13.94	0.8955	5.324	3.379	2.259	4.805	1
4	16.14	14.99	0.9034	5.658	3.562	1.355	5.175	1

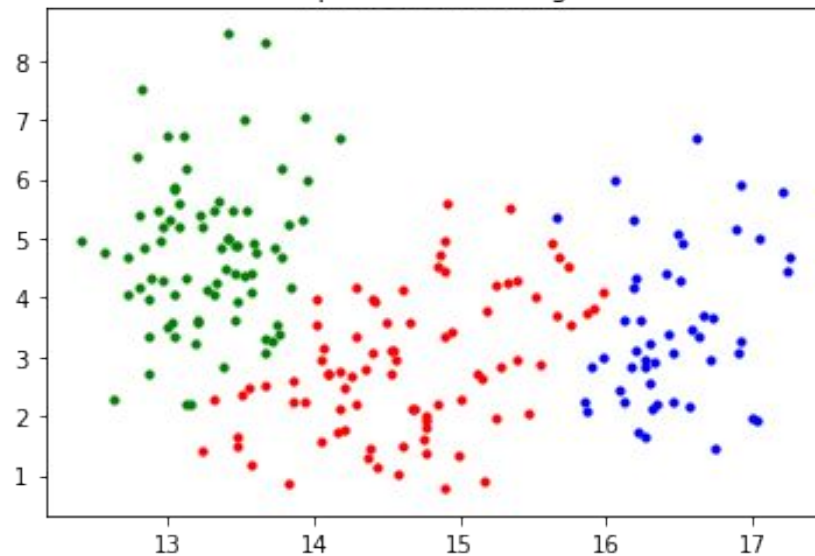
All of these parameters were real-valued continuous.

Seeds dataset

K-means clustering



Spectral Clustering



Conclusion

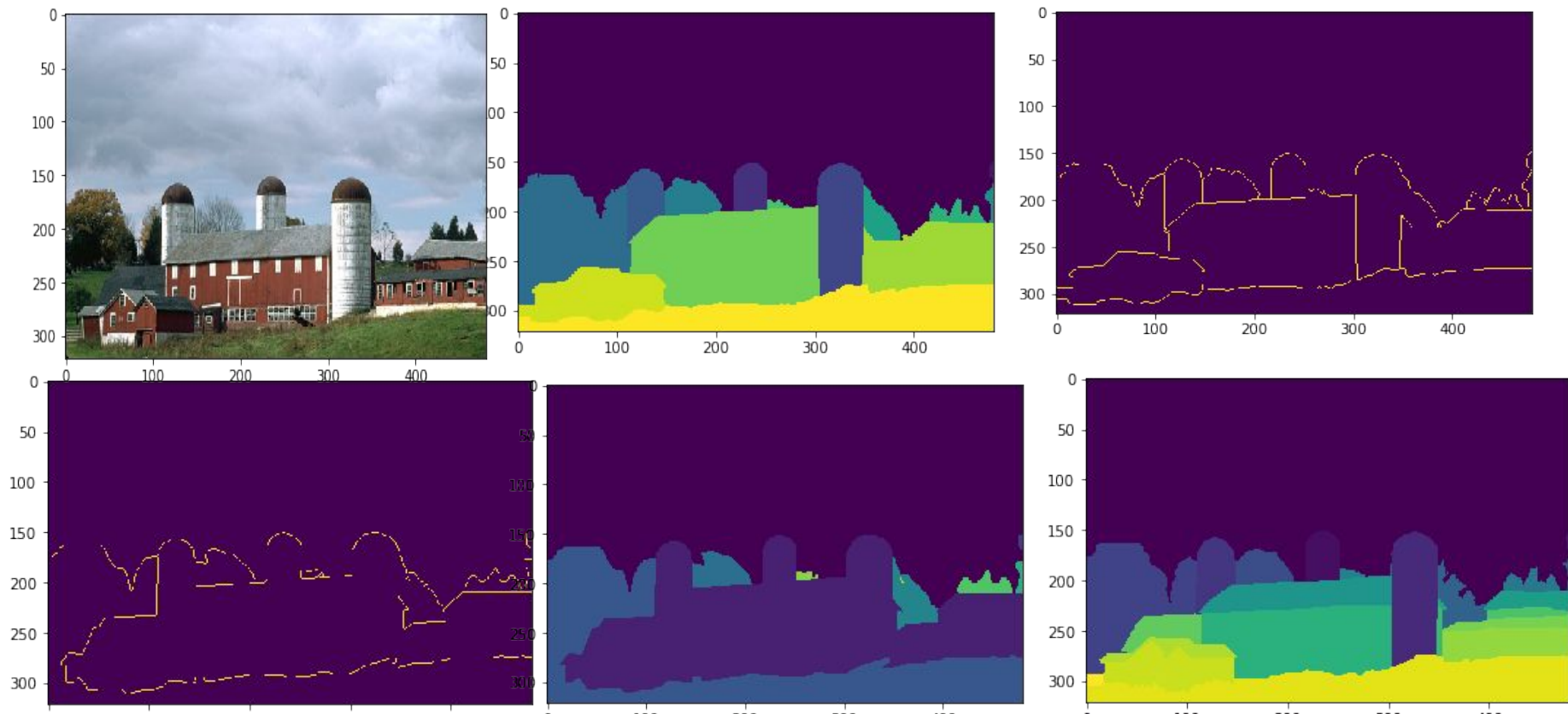
K-Means

- Fast and Simple
- “Embarrassingly parallel”
- Not very useful on anisotropic data

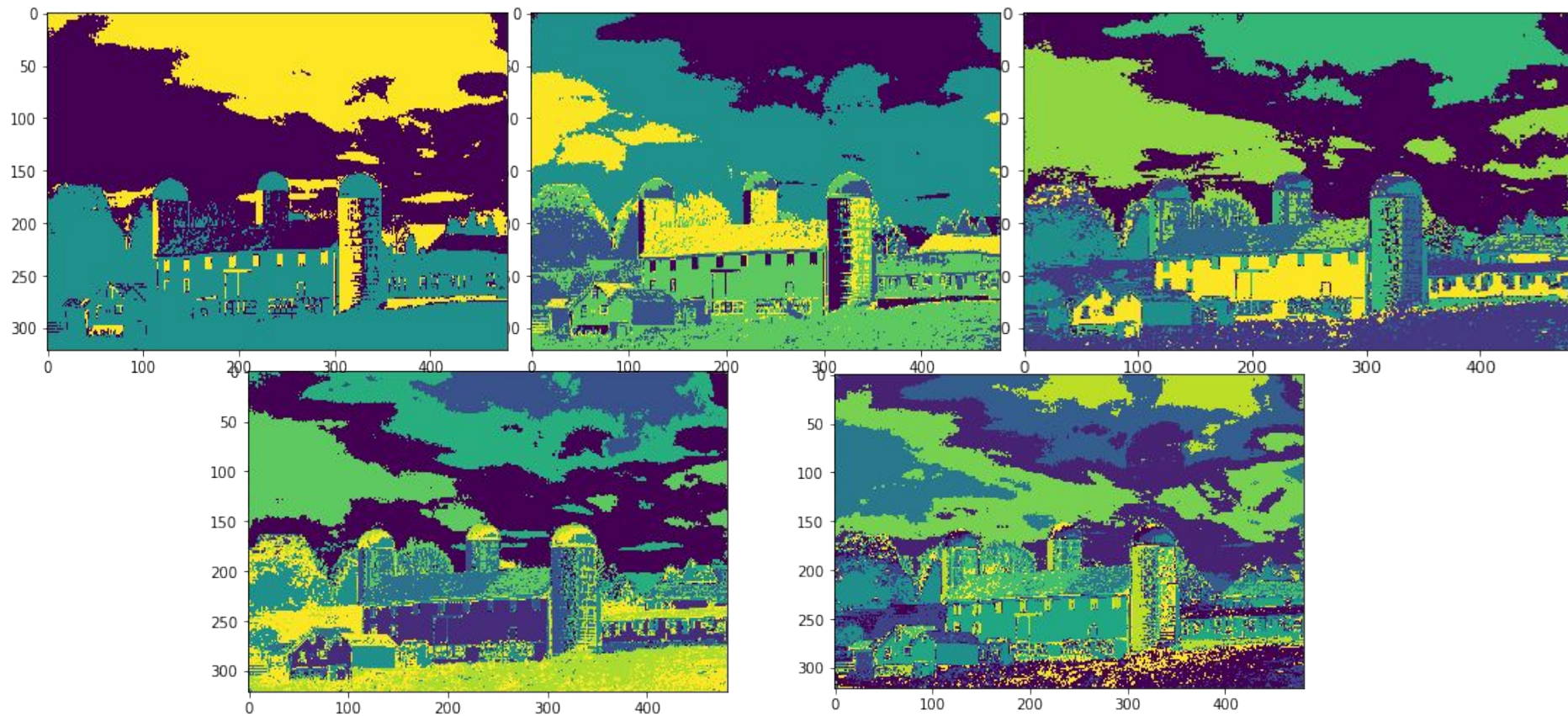
Spectral clustering

- Excellent quality under many different data forms
- Much slower than KMeans

Image segmentation: Original image and GroundTruth



K-means: k=3,5,7,9,11



Spectral clustering: $k=3,5,7,9,11$

