# Digits Clusterization

**Unsupervised Machine Learning**

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# Framing the problem

Trying to find out the best number of clusters to understand out data, we followed the next steps:

1. Dimensionality Reduction: Finding is the best method for our data (SVM Classifier)
2. Clustering Techniques
    - Kmeans
    - MiniBatchKmeans
    - Birch
    - Agglomerative
3. Performance Conclusions: Metrics
4. Next steps: Tuning

# 1. Dimensionality Reduction

# 1. DIMENSIONALITY REDUCTION: Finding the best method for our data

## PCA

(30 components, 85% variance ratio)

| | precision | recall | f1-score | support |
|---|---|---|---|---|
| 0 | 1.00 | 1.00 | 1.00 | 63 |
| 1 | 1.00 | 0.98 | 0.99 | 59 |
| 2 | 0.98 | 1.00 | 0.99 | 55 |
| 3 | 1.00 | 0.99 | 0.99 | 68 |
| 4 | 1.00 | 1.00 | 1.00 | 66 |
| 5 | 0.96 | 0.98 | 0.97 | 52 |
| 6 | 1.00 | 1.00 | 1.00 | 54 |
| 7 | 1.00 | 0.98 | 0.99 | 62 |
| 8 | 0.96 | 0.98 | 0.97 | 51 |
| 9 | 0.95 | 0.95 | 0.95 | 64 |
| accuracy | | | 0.99 | 594 |
| macro avg | 0.99 | 0.99 | 0.99 | 594 |
| weighted avg | 0.99 | 0.99 | 0.99 | 594 |

Accuracy: 0.99 %

?

## LDA

(3 comp, 95% variance ratio)

| | precision | recall | f1-score | support |
|---|---|---|---|---|
| 0 | 1.00 | 0.98 | 0.99 | 53 |
| 1 | 1.00 | 0.98 | 0.99 | 42 |
| 2 | 0.98 | 1.00 | 0.99 | 41 |
| 3 | 1.00 | 0.96 | 0.98 | 52 |
| 4 | 1.00 | 1.00 | 1.00 | 47 |
| 5 | 1.00 | 0.95 | 0.97 | 39 |
| 6 | 0.98 | 1.00 | 0.99 | 43 |
| 7 | 1.00 | 0.96 | 0.98 | 48 |
| 8 | 0.92 | 0.97 | 0.95 | 37 |
| 9 | 0.92 | 1.00 | 0.96 | 48 |
| accuracy | | | 0.98 | 450 |
| macro avg | 0.98 | 0.98 | 0.98 | 450 |
| weighted avg | 0.98 | 0.98 | 0.98 | 450 |

Accuracy: 0.98 %

# 1. DIMENSIONALITY REDUCTION: Finding the best method for our data

## SVM on original data

| | precision | recall | f1-score | support |
|---|---|---|---|---|
| 0 | 1.00 | 0.98 | 0.99 | 63 |
| 1 | 0.95 | 1.00 | 0.98 | 59 |
| 2 | 1.00 | 1.00 | 1.00 | 55 |
| 3 | 1.00 | 0.99 | 0.99 | 68 |
| 4 | 0.99 | 1.00 | 0.99 | 66 |
| 5 | 0.96 | 0.98 | 0.97 | 52 |
| 6 | 1.00 | 1.00 | 1.00 | 54 |
| 7 | 1.00 | 0.98 | 0.99 | 62 |
| 8 | 0.96 | 0.96 | 0.96 | 51 |
| 9 | 0.97 | 0.94 | 0.95 | 64 |
| accuracy | | | 0.98 | 594 |
| macro avg | 0.98 | 0.98 | 0.98 | 594 |
| weighted avg | 0.98 | 0.98 | 0.98 | 594 |

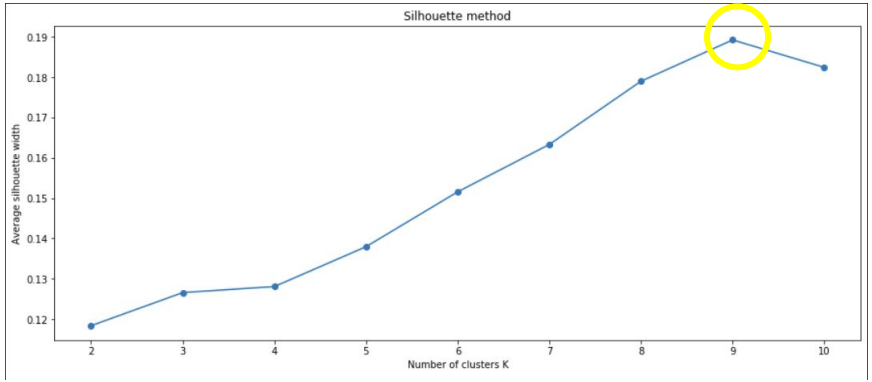
Accuracy: 0.98 %

| | ✓ | ✗ (Under 0.95) | Precision | Recall | F1-score |
|---|---|---|---|---|---|
| Original | 5 | 0 | 98% | 98% | 98% |
| PCA | 6 | 0 | 99% | 99% | 99% |
| LDA | 6 | 2 | 98% | 98% | 98% |

# 2. Clustering Techniques

# 2.0. Checking # clusters

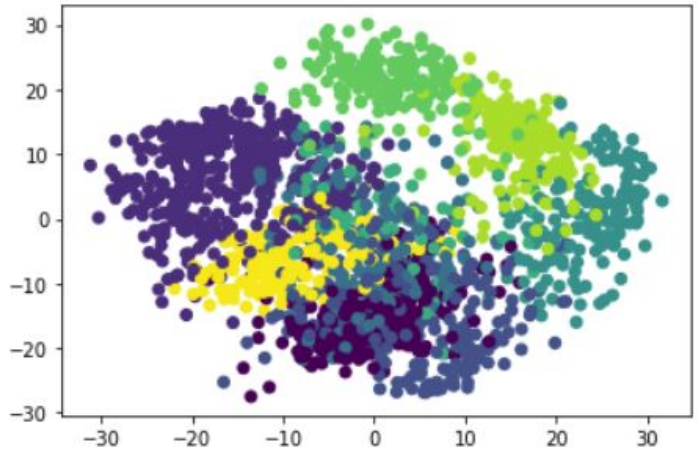
Silhouette Method → 9 clusters




Elbow Method → ?

# 2.1. KMeans

- With PCA




```
Kmeans_model_pca_09
Time execution: 1269.8
Homogeneity: 0.692
Completeness: 0.750
V-measure: 0.719
Adjusted Rand Index: 0.597
Adjusted Mutual Information: 0.717
==========================
```

# 2.2. MiniBatchKMeans (on PCA & 9 clust)

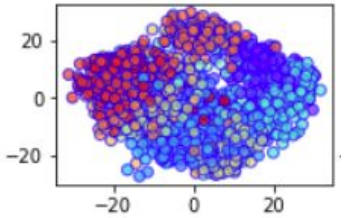
- We expected quite faster results that with KMeans but in this case the difference has just slightly improved:

1269.8 vs 1176.04

```
============================
Mini-batch Kmeans
Time execution: 1176.04
Homogeneity: 0.252
Completeness: 0.531
V-measure: 0.342
Adjusted Rand Index: 0.196
Adjusted Mutual Information: 0.340
============================
```

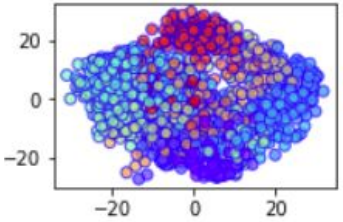
# 2.3. Birch

brc_02 = Birch(branching_factor=2, n_clusters=None, threshold=1.5)



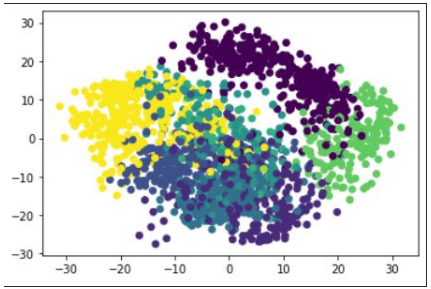

```
Birch_02
Time execution: 3100.32
Homogeneity: 1.000
Completeness: 0.307
V-measure: 0.470
Adjusted Rand Index: 0.000
Adjusted Mutual Information: -0.000
=========================
Birch_50
Time execution: 3091.55
Homogeneity: 1.000
Completeness: 0.307
V-measure: 0.470
Adjusted Rand Index: 0.000
Adjusted Mutual Information: -0.000
```

brc_50 = Birch(branching_factor=50, n_clusters=None, threshold=1.5)



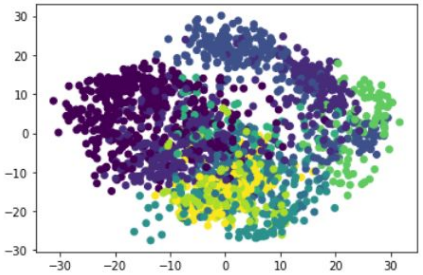
# 2.4. Agglomerative

Linkage = **average**



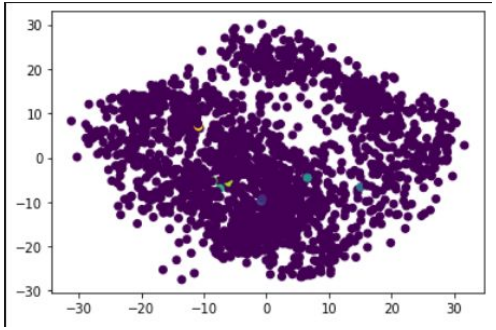
```
============================
Agg_complete_linkage
Time execution: 19.85
Homogeneity: 0.515
Completeness: 0.593
V-measure: 0.551
Adjusted Rand Index: 0.361
Adjusted Mutual Information: 0.547
Silhouette Coefficient: 0.101
============================
============================
Agg_average_linkage
Time execution: 397.57
Homogeneity: 0.697
Completeness: 0.804
V-measure: 0.747
Adjusted Rand Index: 0.601
Adjusted Mutual Information: 0.745
Silhouette Coefficient: 0.158
============================
```

Linkage = **complete**



# 2.4. Agglomerative

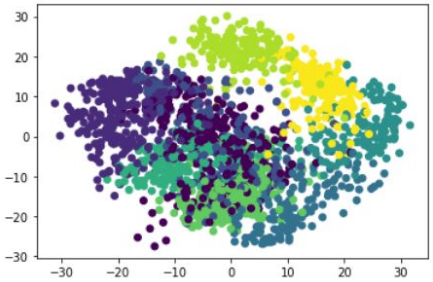
Linkage = **single**




```
============================
Agg_single_linkage
Time execution: 10.41
Homogeneity: 0.005
Completeness: 0.275
V-measure: 0.009
Adjusted Rand Index: 0.000
Adjusted Mutual Information: 0.000
Silhouette Coefficient: -0.136
============================
```

```
============================
Agg_ward_linkage
Time execution: 11.04
Homogeneity: 0.772
Completeness: 0.821
V-measure: 0.796
Adjusted Rand Index: 0.691
Adjusted Mutual Information: 0.794
Silhouette Coefficient: 0.169
============================
```

Linkage = **ward**



# 2. Clustering Conclusions

KMeans, MiniBatchKMeans, Agglomerative, Birch

- **KMeans PCA over KMeans**:

  ↓ Computational cost

- **MiniBatchKmeans over KMeansPCA**:

  ↓ Time
  ↓ Performance

- **Agglomerative**:

  Different linkages perform indistinctly mediocre except for the “single” linkage which is even worse because results in a single cluster.

- **Birch:**

  It has the best performance in Time, Homogeneity, V-measure, and David Bouldin metrics. Completeness is quite good too.

# 3. Performance Conclusions: METRICS

# 3. PERFORMANCE CONCLUSIONS: Metrics

| METRIC | TECHNIQUE | SCORE 1 | TECH 2 | TECH 3 |
|---|---|---|---|---|
| *Time execution:* | OPTICS | 1123.58 | DB-SCAN | Birch_50 |
| *Homogeneity:* | Birch_02, Birch_50 & DBSCAN | 1 | Kmeans | Agg_avg_link |
| *Completeness:* | OPTICS | 1 | Agg_avg_link | Kmeans |
| *V-measure:* | Kmeans | 0.71 | Agg_avg_link | Birch_02 & Birch_50 |
| *Adjusted Rand Index:* | Kmeans | 0.59 | Agg_avg_link | MiniBatch |
| *Adjusted Mutual Information:* | Kmeans | 0.71 | Agg_avg_link | MiniBatch |


Kmeans
(default n_clusters = 9)
2
1
3

Gràcies!

¡Gracias!

Thank you!