

OPM 562

Case study:

Unsupervised learning for
failure mode detection

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04.05.2020





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1. Data cleaning & application of ML methods

Data preparation

Before further analysis, 24 rows were dropped from the dataset:

- 19 rows with missing data
- 5 rows with negative or zero values

To get deeper insights into the features, the dataset was divided into 2 parts:

- Frame data (top tube, seat tube, chainstay, front center)
- Battery data (weight, voltage, capacity)

ML methods

Combination of PCA and K-Means clustering to improve segmentation results:

- Dimension reduction prior to data segmentation
- PCA helps reduce noise and multicollinearity, making features both statistically significant as well as non-correlated with each other
- K-Means clustering method uses Euclidean distance as measurement of within cluster distance but does not work well in high dimensions
- We reduce the number of dimensions first so that K-means method produces more meaningful results

2.1 Frame Analysis

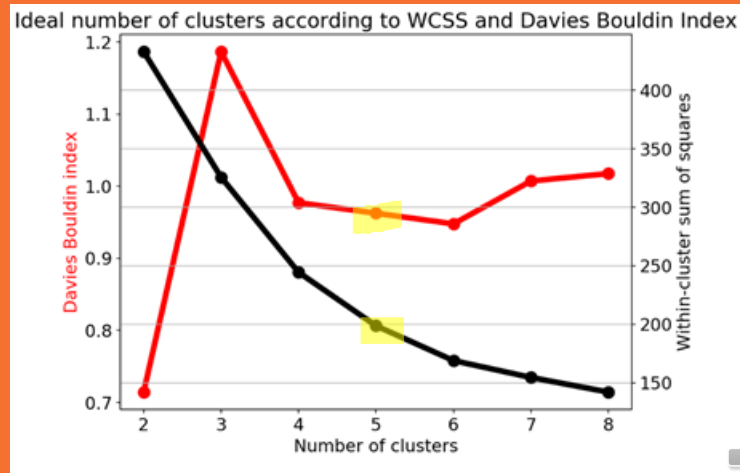
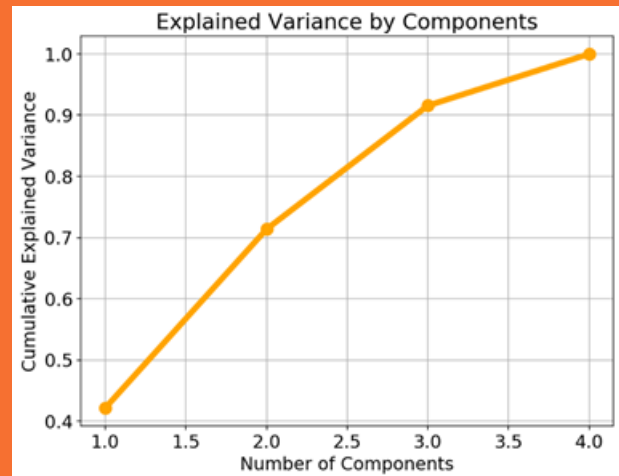
2.1.1 PCA & K-Means

Principal Component Analysis

- Original number of components: 4
- By reducing to 3 components, we are still able to explain 90% of the variance
- Thus, adjusted number of components: 3

K-Means Clustering

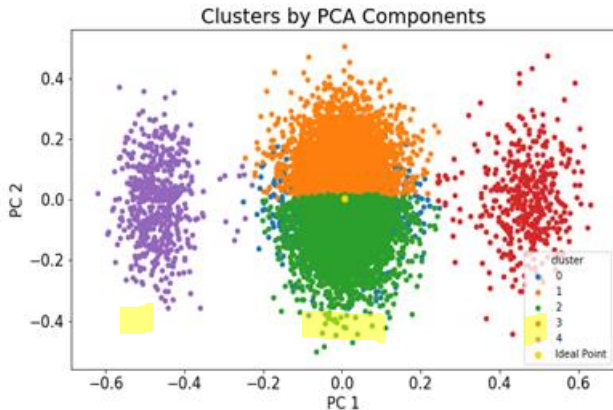
- By looking at the Elbow graph (black), it can be noted that the point of inflection occurs approximately around the 5th cluster.
- By choosing 5 clusters, we get a Davies Bouldin Index of 0.9476
- Thus, number of clusters: 5



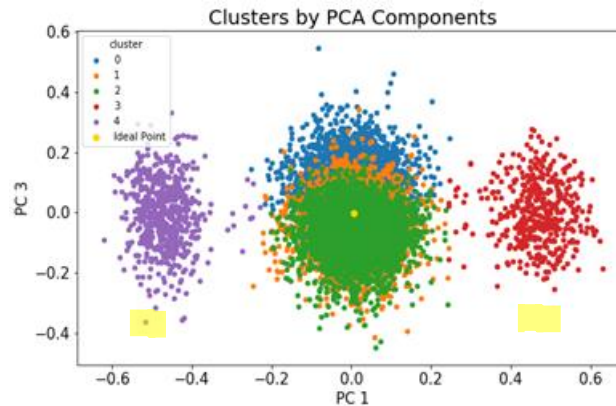
2.1 Frame Analysis

2.1.2 Clustering

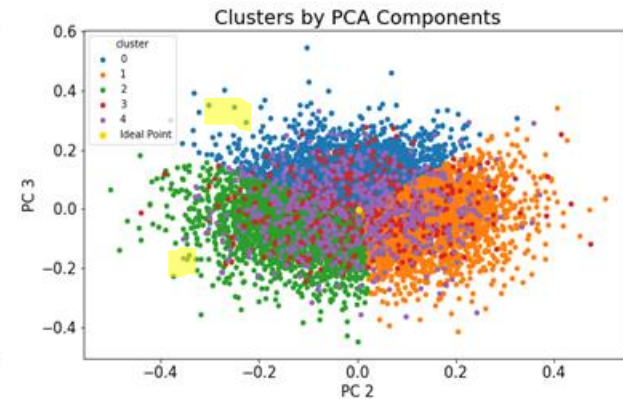
Bike frame with **ideal** features (nominal values) belongs to **cluster 0**



- **Clusters 3 and 4** clearly separated with respect to **PC 1**
- **PC 1** is mainly explained by **top tube** and **front center**



- **Clusters 1 and 2** determined by **PC 2**
- **PC 2** is mainly explained by **chainstay**



- **Clusters 0 and 2** determined by **PC 3**
- **PC 3** is mainly explained by **seat tube**



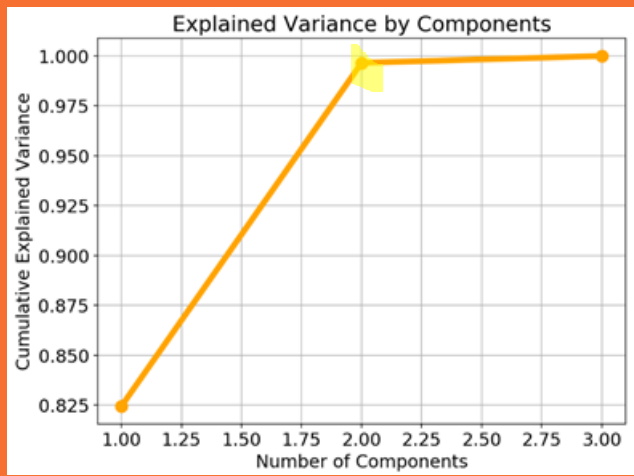
2.1 Frame Analysis

2.1.3 Deviation from nominal values

- Low deviations from nominal value
- Medium-sized deviations
- Large deviations

	Nominal value (in mm)	Cluster 0 28.24%	Cluster 1 31.78%	Cluster 2 30.2%	Cluster 3 4.3%	Cluster 4 5.48%
Top tube	602	601.998879 -0,001121	602.021299 +0,021299	601.994177 -0,005823	600.234905 -1,765095	604.624928 +2,624928
Seat tube	584	584.962837 +0,962837	583.602228 -0,397772	583.555549 -0,444451	583.922130 -0,077870	584.051143 +0,051143
Chainstay	487	486.889788 -0,110212	487.935831 +0,935831	486.045182 -0,954818	486.940714 -0,059286	486.970386 -0,029614
Front center	633	632.988222 -0,011778	632.997101 -0,002899	632.991718 -0,008282	636.775588 +3,775588	630.257684 -2,742316





2.2 Battery Analysis

2.2.1 PCA & K-Means

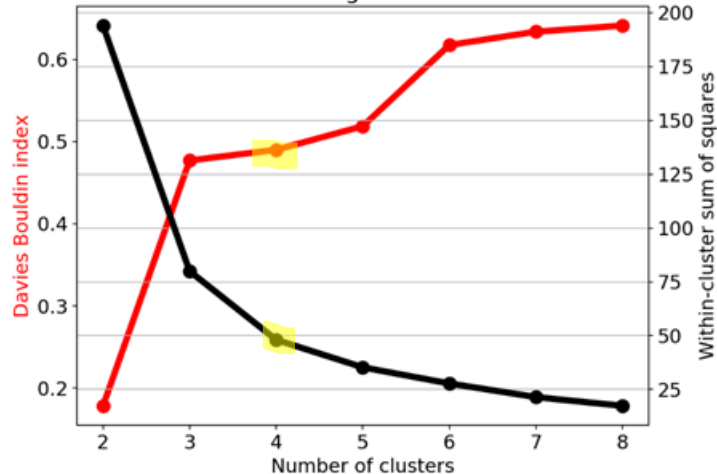
Principal Component Analysis

- Original number of components: 3
- By reducing to 2 components, we are still able to explain almost 100% of the variance
- Thus, adjusted number of components: 2

K-Means Clustering

- By looking at the Elbow graph (black), it can be noted that the point of inflection occurs approximately around the 4th cluster.
- By choosing 4 clusters, we get a Davies Bouldin Index of 0.5188

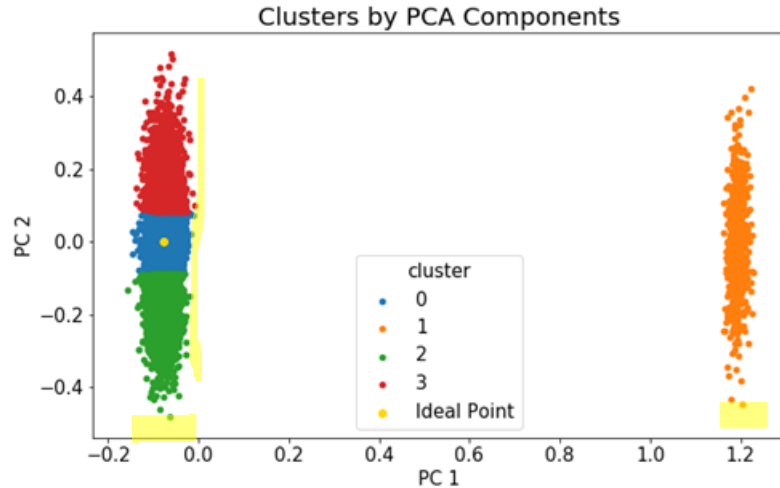
Ideal number of clusters according to WCSS and Davies Bouldin Index



2.2 Battery Analysis

2.2.2 Clustering

- Battery with **ideal** features (nominal values) belongs to **cluster 0**
- Two clearly separated clusters with respect to PC 1
- Clusters on the left around the ideal point, cluster on the right much higher than ideal point.
- **PC1** is mainly explained by **voltage** and **capacity**
- Clusters 0, 2 and 3 differ with respect to PC 2
- PC 2 is mainly explained by battery weight



2.1 Battery Analysis

2.2.3 Deviation from nominal values

- Low deviations from nominal value
- Medium-sized deviations
- Large deviations

	Nominal value	Cluster 0 44.2%	Cluster 1 6.04%	Cluster 2 24.16%	Cluster 3 25.59%
Battery weight	2.8 kg	2.800075 kg + 0.00075 kg	2.799693 kg - 0.000307 kg	2.805238 kg + 0.005238 kg	2.794915 kg - 0.005085 kg
Full charge voltage	400 Wh	400.006 Wh + 0.006 Wh	281.245 Wh -118.755 Wh	400.002 Wh + 0.002 Wh	399.982 Wh - 0.018 Wh
Battery capacity	36 V	36.000318 V + 0.000318 V	34.198396 V - 1.801604 V	36.001027 V + 0.001027 V	35.998603 V - 0.001397 V



3. Managerial implications

Quality of frames

Reasonable:
28% of frames (cl. 0)

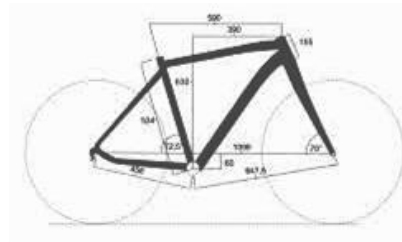
Questionable:
62% of frames (cl. 1&2)

Poor:
10% of frames (cl. 3&4)

Acceptable deviations from the nominal values, except for rather long seat tube

Somewhat shorter seat tube & noticeably longer or shorter chainstay

Very short top tube and very long front center or vice versa



Top tube and front center are inversely proportional, which in particular leads to 2 clusters of poor quality.

From the bike scheme, it seems like these deficiencies result from inaccurate connection of the tubes. It can result from not properly set up machines and/or too rare maintenance.



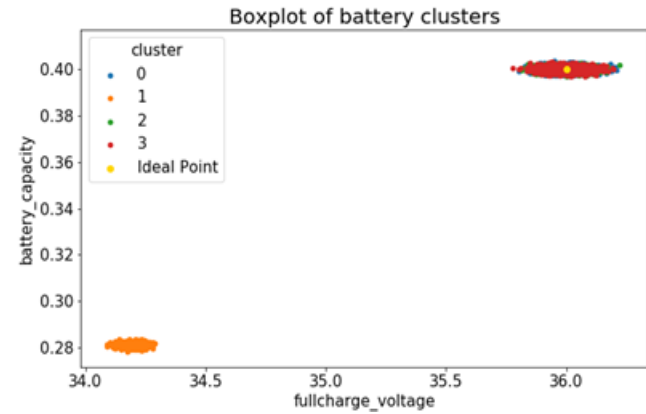
3. Managerial implications

Quality of batteries

44% batteries (cl. 0) - **good** quality with acceptable deviations from nominal values.

50% batteries (cl. 2&3) - **reasonable** quality with rather low/high weight.

6% batteries (cl. 1) - **poor** quality with very low capacity and very low full charge voltage.



Overall conclusions and suggestions

- **27%** of bikes - good or reasonable quality (over both parts)
- **69%** of bikes have issues with either frame or battery
- **4%** of questionable or poor quality bikes (over both parts)

Parts produced by Sunflower Bikes

Analyse machines data, investigate causes of defects

Adjust schedules of maintenance, improve accuracy of machines

Parts bought from a supplier

Communicate the problem to the supplier

Implement a stricter sampling policy, reject shipments with high % of defects

Thank you

