MotoGP

Circuit clustering analysis

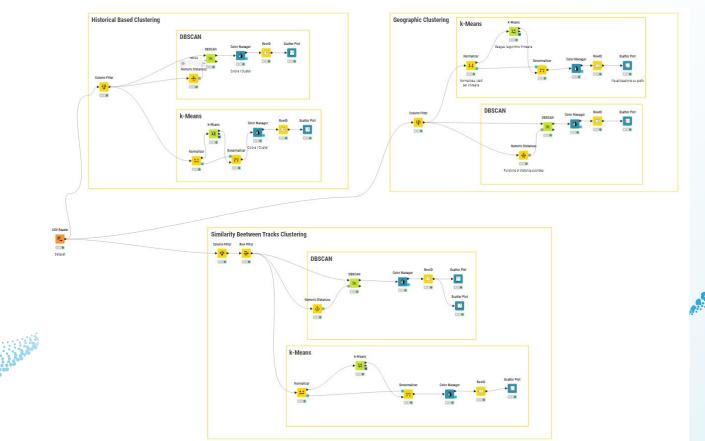
Introduction

- Dataset
- KNIME Workflow
- Geographical clustering
- Circuit similarity clustering
- Historical clustering
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- Resources

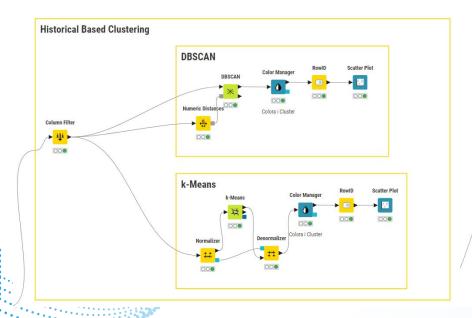
Dataset

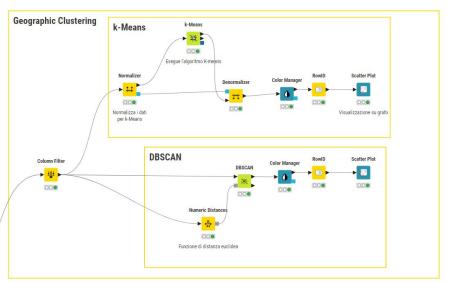
- The dataset builds upon "MotoGP circuits" by Mike Enting on Kaggle.
- It has been **modified** to add a "Short Name" column to improve the readability of the plots when hovering the mouse.
- The coordinates have been **split** into two columns to <u>improve</u> usage of the dataset.

KNIME Workflow

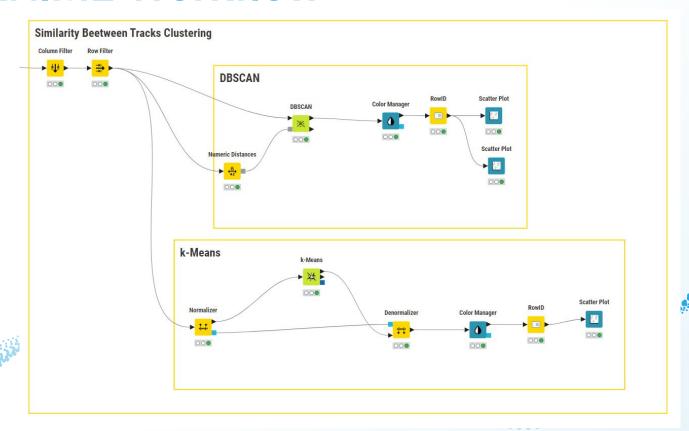


KNIME Workflow



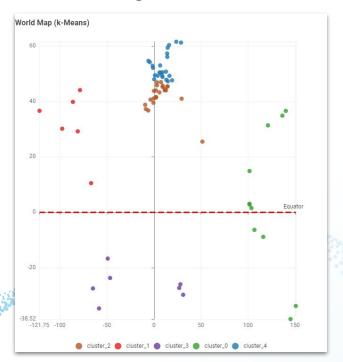


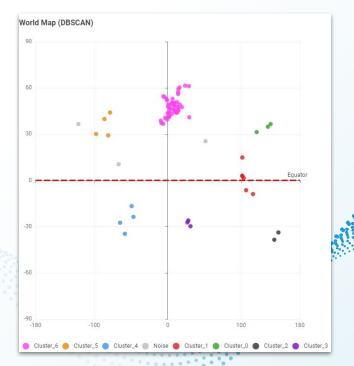
KNIME Workflow



Geographic Clustering

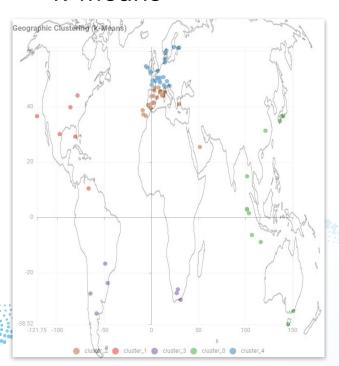
Clustering based on coordinates





Geographic Clustering

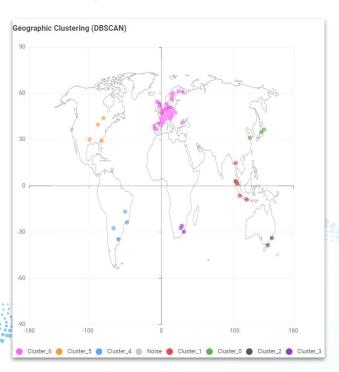
k-Means



k-Means uses a normalization function that has **altered** the scale of the earth, but still found the different clusters: *North and South of Europe, North America, South America (with South Africa) and Asia (with Oceania).*

Geographic Clustering

DBSCAN



The density based algorithm has identified the different continents. There are **3 outliers** in the model: Lusail(Qatar), Laguna Seca (USA), San Carlos (Venezuela). They are too isolated and cannot be associated with any other cluster.

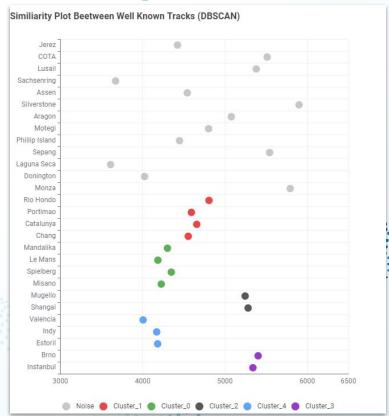
This clustering takes into consideration a couple of characteristics of the tracks (only well known tracks have been selected):

- N° of Left & Right Turns
- Avg Width
- Length
- Longest Straight

DBSCAN

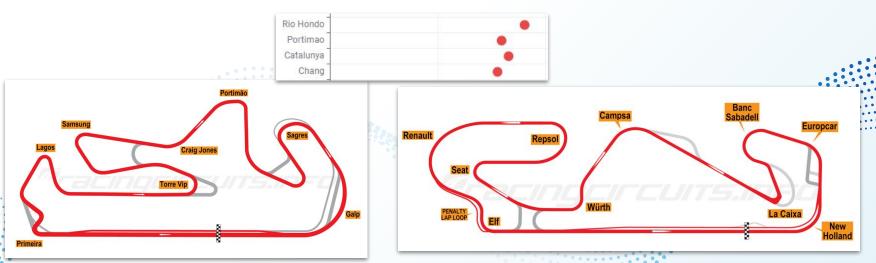
As expected, a lot of **outliers** are present.

The **epsilon** and the **min. Points** parameters have been adjusted to have some reasonable clusters.



Example: Portimao-Catalunya

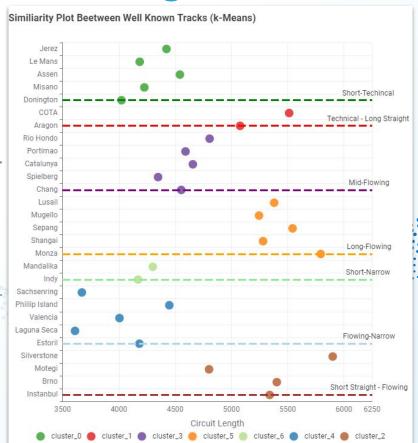
We can use this example to show how the model works. Portimao and Catalunya are often compared and are considered to be very similar tracks. The algorithm has correctly put them together.



k-Means

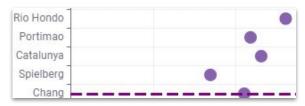
The tweaking of the **"k"** parameter (for the number of clusters) has been applied here to better understand the results.

Since k-Means does **not detect outliers**, and clusters the whole dataset, captions have been added manually to have a better comprehension of the clusters.



A closer look k-Means

Portimao and **Catalunya** are still best friends.





Misano and Jerez is another great example.

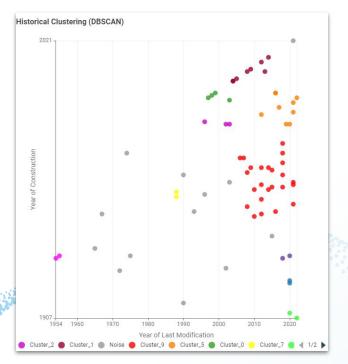


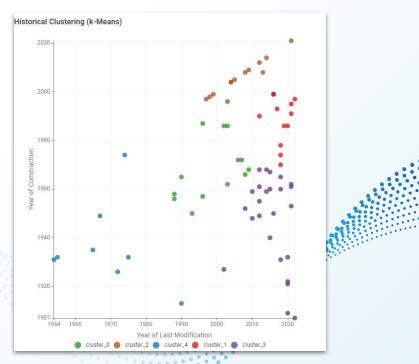




Historical Clustering

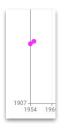
Clustering based on construction and modification year





Historical Clustering

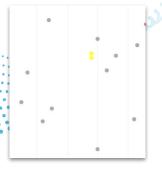
DBSCAN highlights



Berne(1931,1954) and Belfast(1932,1955) are both old and now closed and unused tracks.



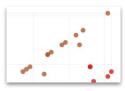
These are tracks that are more recent, like **Shanghai** (2004), **Istanbul** (2005), **COTA** (2012), **Chang** (2014), that have been built recently and have not been changed since. The **outlier** (in gray) is **Mandalika** (2021).



There is a lot of noise in this dataset (and not dense enough), and a **lot of tracks** have not been included in any cluster.

Historical Clustering

k-Means highlights



The brown cluster has the recent, less changed tracks, **Mandalika (2021)** has been included in the cluster.



This green cluster contains older tracks, still used today, like Suzuka(1962,2003), Misano(1972,2007) and Le Mans (1966-2008) that have not been changed in recent years.



These are Indianapolis(1909,2020) and Island of Man (1907,2022) the two oldest in the dataset. In k-Means they have been added to the larger purple cluster. However this is not very accurate, and only DBSCAN highlights their uniqueness.

Possible improvements

- Adding more data to the circuits:
 - Maximum Elevation change
 - Anti/Clockwise
 - Asphalt type/Kerb height/Altitude
 - Lap Record/Record Holder/Top Speed/Min Speed
- Changing distance functions
- Changing clustering algorithm parameters

Resources

- Original, pre-modification, Dataset: "MotoGP circuits" by Mike Enning on Kaggle
- Track layout pictures by <u>RacingCircuits.info</u>
- KNIME
- To access all the resources please visit the <u>GitHub repository</u>

The End