

MachineLearnAthon – Feature Engineering

Recorded by Lara Kuhlmann



Learning outcomes of today

After successfully completing this micro-lecture, you are able to....

- Enrich datasets with external information
- Extract informative features from datetime values
- Merge and aggregate multiple data sources
- Apply normalization and logarithmic transformations









Agenda for today

- **Data Enrichment**
- **Datetime Feature Extraction**
- Combining datasets merge and aggregation
- Data normalization and Log-Transformation











Data Enrichment

- Refers to the process of enhancing, refining or organizing data to extract valuable insight from it. [1]
- The purpose of data enrichment is to discern relationships, clusters, semantic ontologies
 within a collection of data that unveil new insights to make informed decisions. [2]

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- Benefits of data enrichment: [1]
 - Diagnosis and mitigation of potential risks
 - Prompt identification of new opportunities
- Example:
 - Weather or temperature data for energy or sales forecasting
 - Currency exchange rates or traffic intensity









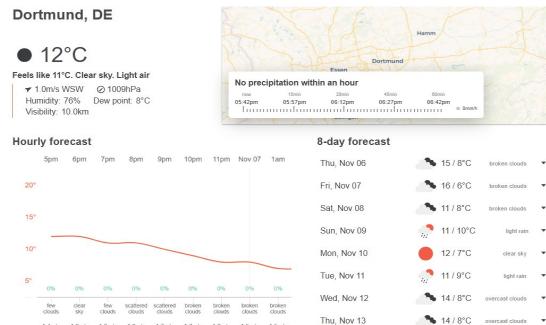


Data Enrichment Example – Weather APIs

- External weather data can inject real-world signal into predictive models.
- Online sources:
 - Open-Meteo: Global weather model data, hourly or daily resolution, very simple API-Calls, no API key required.

OpenWeatherMap: Global coverage, historic data and forecast, granularity down to hourly,

API key required











Datetime Feature Extraction

- Datetime features carry seasonal or temporal patterns useful for Machine Learning models.
- DatetimeFeatures() automatically extracts several date and time features from datetime variables.
- DatetimeFeatures() has:
 - Extract time features
 - Extract date and time features
 - Time series
- Derived features can include:
 - Month, quarter, year
 - Weekday/weekend indicator
 - Hour of the day





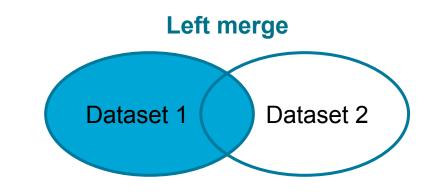






Combining datasets – Merge (I)

- If the datasets have at least variable in column, they can be used as keys to join the datasets
- The merge can either be left, right, inner or outer
- In case of a left merge, only the keys from the left dataset are used
- This may result in missing values in the column from the right dataset



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Source: https://pandas.pydata.org/docs/user_guide/merging.html

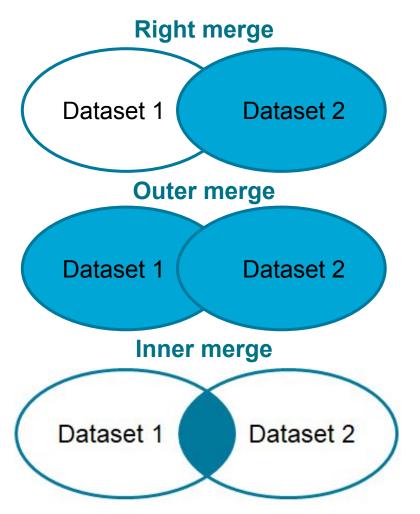




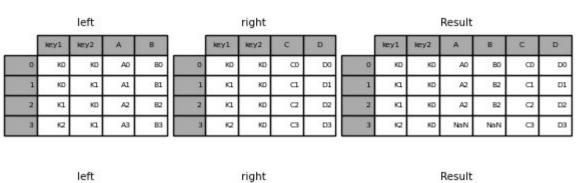




Combining datasets – Merge (II)



Source: https://pandas.pydata.org/docs/user_guide/merging.html



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Combining datasets – Aggregation (I)

- Aggregation is a GroupBy operation that reduces the dimension of the grouping object.
- The result of an aggregation is a scalar value for each column in a group.

• Example:

	Animal	Height	Weight
0	Cat	9.1	7.9
1	Dog	6.0	7.5
2	Cat	9.5	9.9
3	Dog	34.0	198.0

animals.groupby("Animal").sum() has output:

Animal	Height	Weight
Cat	18.6	17.8
Dog	40.0	205.5

Source: https://pandas.pydata.org/docs/user_guide/groupby.html











Combining datasets – Aggregation (II)

- Built-In Aggregation methods include:
 - Any() compute whether any of the values in the groups are truthy
 - All() Compute whether all of the values in the groups are truthy
 - First() Compute the first occuring value in each group
 - Max() Compute the maximum value in each group











Transformation: Data normalization

- Data normalization is a technique used to rescale numerical data into a standardized range, typically between 0 and 1 or -1 and 1
- This ensures that all features contribute equally to analysis or modeling, regardless of their original scale
- The most commonly used normalization technique is the Min-Max Scaling

$$X_{normalized} = \frac{X - X_{min}}{X_{max} - X_{min}}$$

García et al. (2015): Data Preprocessing in Data Mining, p.46 f.



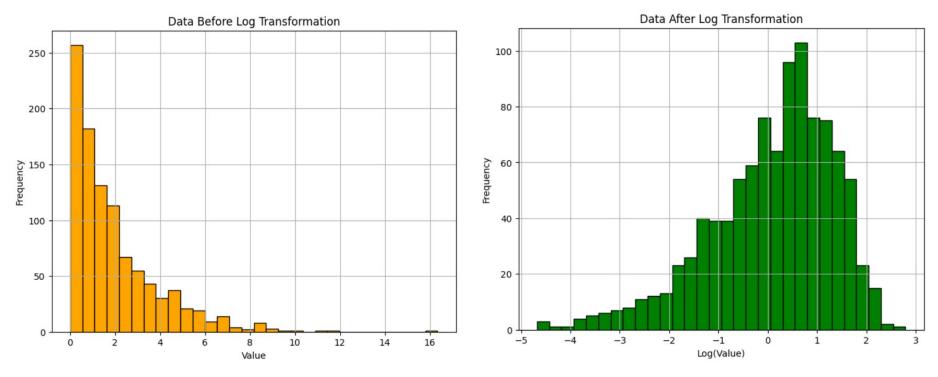






Transformation: Data logarithmic transformation (I)

The log-transformation is widely used to deal with skewed data.



After log-transformation, the data is much more closer to a normal distribution.

Source: https://www.geeksforgeeks.org/data-science/log-transformation/











Recap this lecture

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