DATA-DRIVEN SEARCH FOR TRAFFIC DRIVERS WITH DATA PROVIDED BY



### We are an interdisciplinary team with diverse background in business, science, music and IT

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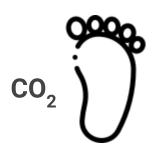
IT Engineer

# We'll show how to create impactful content

data driven and backed by machine learning

- 1. Client & Problem description
- 2. Data & Target
- 3. Insights & Hypothesis
- 4. Modelling
  - a. Baseline
  - b. Feature Engineering
  - c. Predicting Model
- 5. Recommendations
- 6. Future work

### We help EFAHRER in empowering their users to contribute to carbon reduction



**EFAHRER.com is a media portal** which strives to influence users to take actions that support CO<sub>2</sub> reduction.





We want to **provide valuable insights** for the

editorial team



We want to perform a prognosis of article success



### For that we analyzed one of EFAHRER's biggest traffic sources for news articles and enriched the data



Editorial data 6.899 unique articles



Feed Click



**Millions** 

# Data quality and completeness of raw data led to extensive preprocessing and analysis for modelling

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#### Challenges

- Missing values
- Lack of article versions
- 3 different aggregation levels for selected metrics

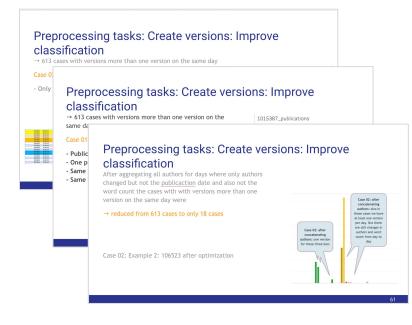
#### Strategies:



Delete







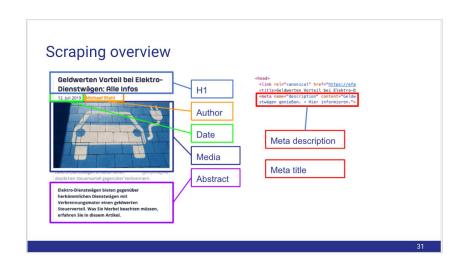
### Scraping of 6 000 articles increased the data quality and added new features



Respecting ethics in web scraping we managed to add

- + 6 visible features
- + 5 meta features (invisible)







### By importing related search terms for 17 product categories we added a trends score to our data



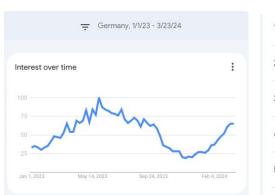
Using a NLP classifier we matched each article with a related search label and trend score



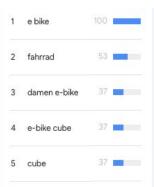
+ 2 features



Search trend for "E-Bike"



Related queries





#### We identified the following relevant features

- Article genre and topic
- Type of the first media: video or image
- ➡ Word count and lengths of the metadata



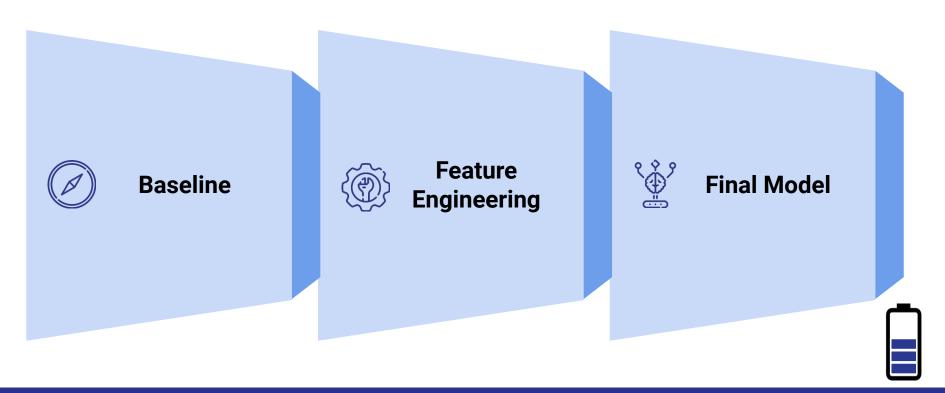


Features in the positive feedback loop with the target were **ignored** or **normalized**:

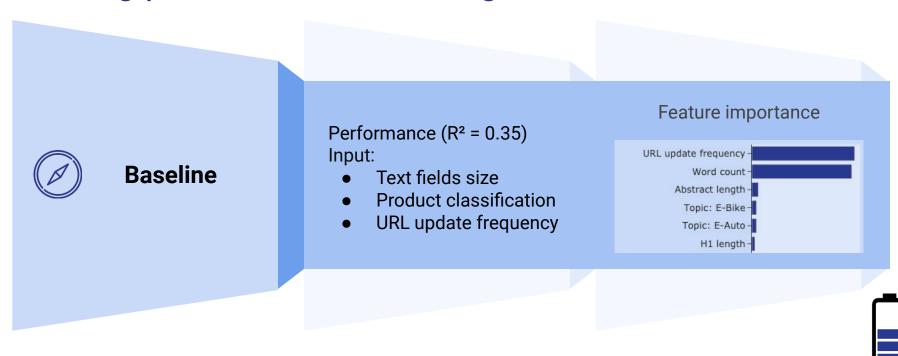
- Number of likes, dislikes, video views
- Number of URLs → URL update frequency



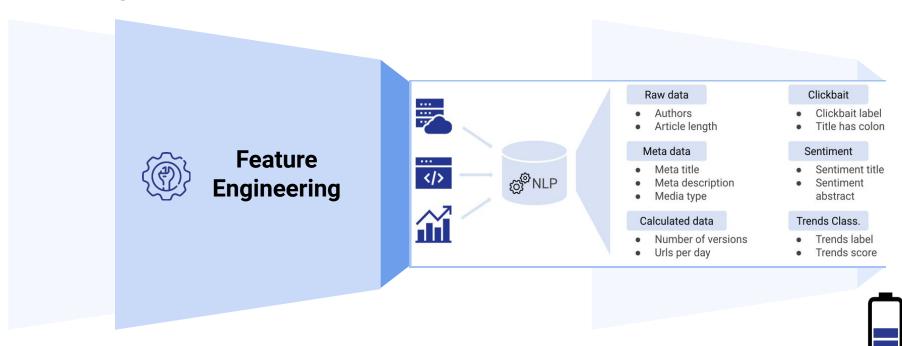
#### We verified our hypotheses and created prediction tool



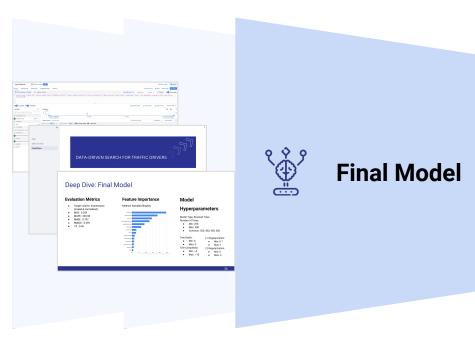
### With the baseline model we created an advanced starting point for our modeling



## We engineered additional features based on the existing data



## Our final model is a stable starting point for predicting article performance



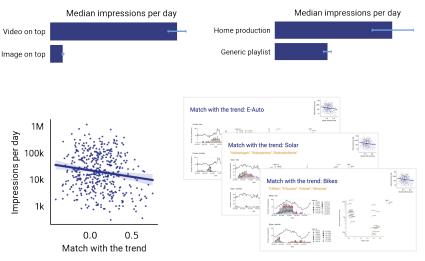
- Modeling with AutoML Tables by Google Vertex AI
- Best performance (R<sup>2</sup> = 0.49) for simpler model without full text features

Live demo



### Updating of articles enhances outcomes, the media plays relevant role

- 1 Change in URL has tangible impact on impressions
- The algorithm prefers articles with videos over images as the first media on page
- 3 Video production pays off!
- 4 Optimize publication timing alongside trends
- 5 Algorithm does not punish clickbait behavior





## Further improvements promise a reliable prediction of page impressions

- Try out different semantic segmentation and model each segment individually (e.g. News)
- Dive deeper into the video and image content and formats
- Refine the trend-related features (e.g. different keyword & time matching, trend sources)
- Improve the evaluation of "clickbaitness"
- Fine tune sentiment analysis
- The full article history would provide new valuable features



### Thank you for your attention

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- Jin-Ho Lee... and others



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- Markus Höllmüller
- Analytics team