



Kickstarter Project Analysis

Machine Learning Project
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What is **KICKSTARTER** ?

- Kickstarter is a funding platform for creative projects
 - allows creators to present a project idea to a broad audience
 - everybody interested in a project can pledge money (backer)
- if the funding goal is reached,
 - the creators get the pledge money
 - and the project is realized
- examples of Kickstarter projects are:
 - Pebble Time, Coolest Cooler, Legend of Vox Machina



kickstarter.com

How does Kickstarter make money?

- Kickstarter charges a 5% fee of the funds raised
 - when a project is successfully funded
- our job is to help creators realize projects
 - more successful creators means higher revenue for Kickstarter
- increase in creator satisfaction
 - higher success rate of their projects
 - less time wasted on unsuccessful projects



kickstarter.com

The Data Set

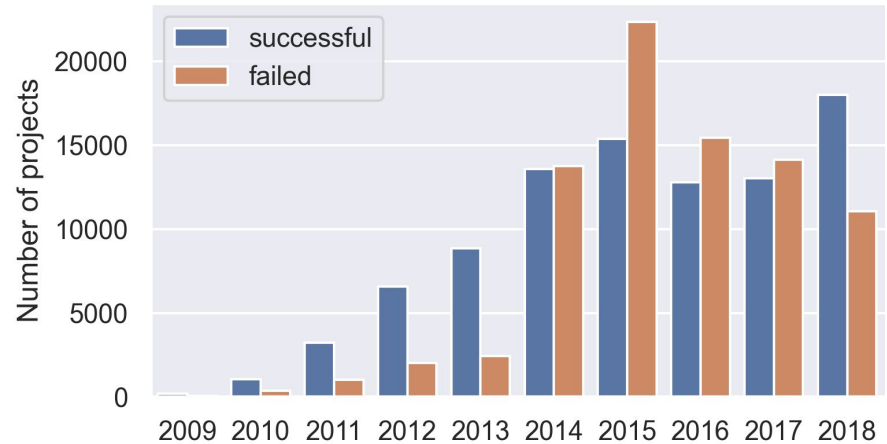
- data from 2009 to 2019
 - only for the first few months of 2019
- distributed over 55 csv files
- a total of 180000 projects
- large amount different features
 - mostly categorical
 - like currencies, text, unix dates, ...



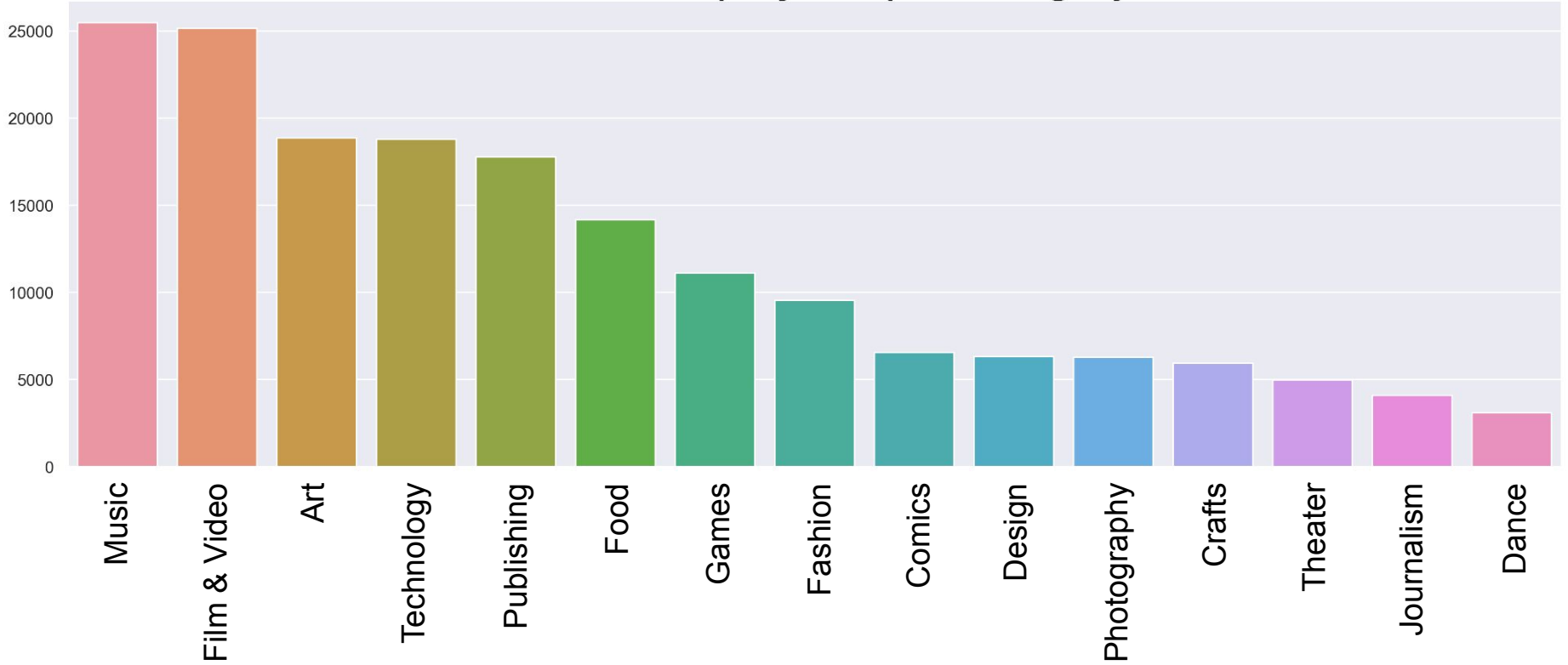
kickstarter.com

What makes a project successful?

- What are popular categories?
- Influence of the funding goal?
- Influence of the length of the funding period?
- Which other factors are important?

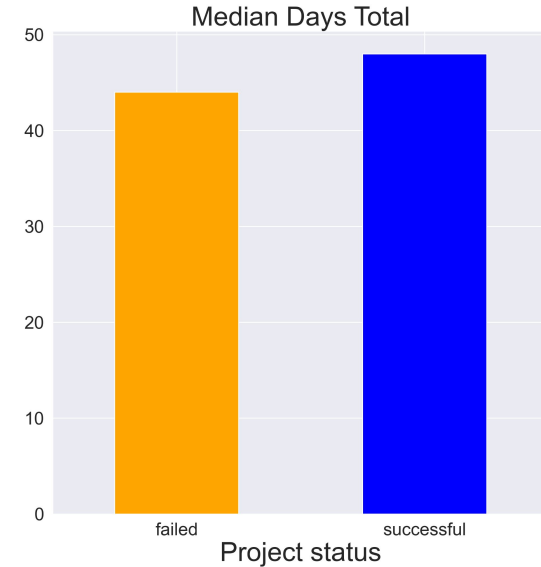


Number of projects per category



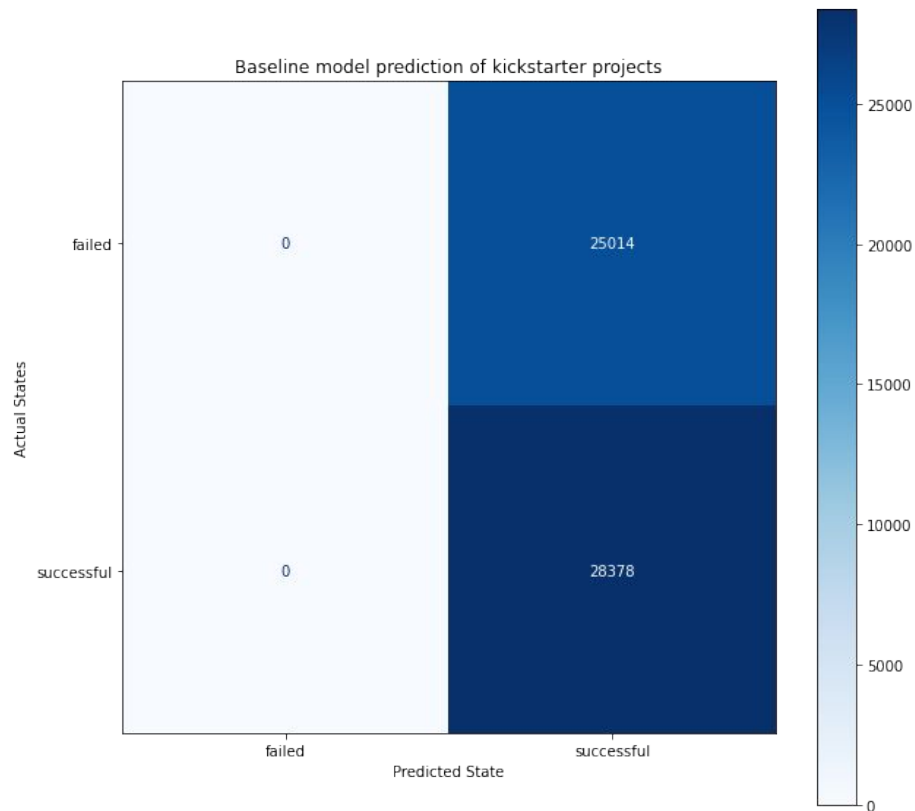
Other factors for project success?

- funding goals
 - high funding goals = more failure
 - lower goals = more success
- duration of the project
 - longer projects are more successful
 - difference is small (effect too?)



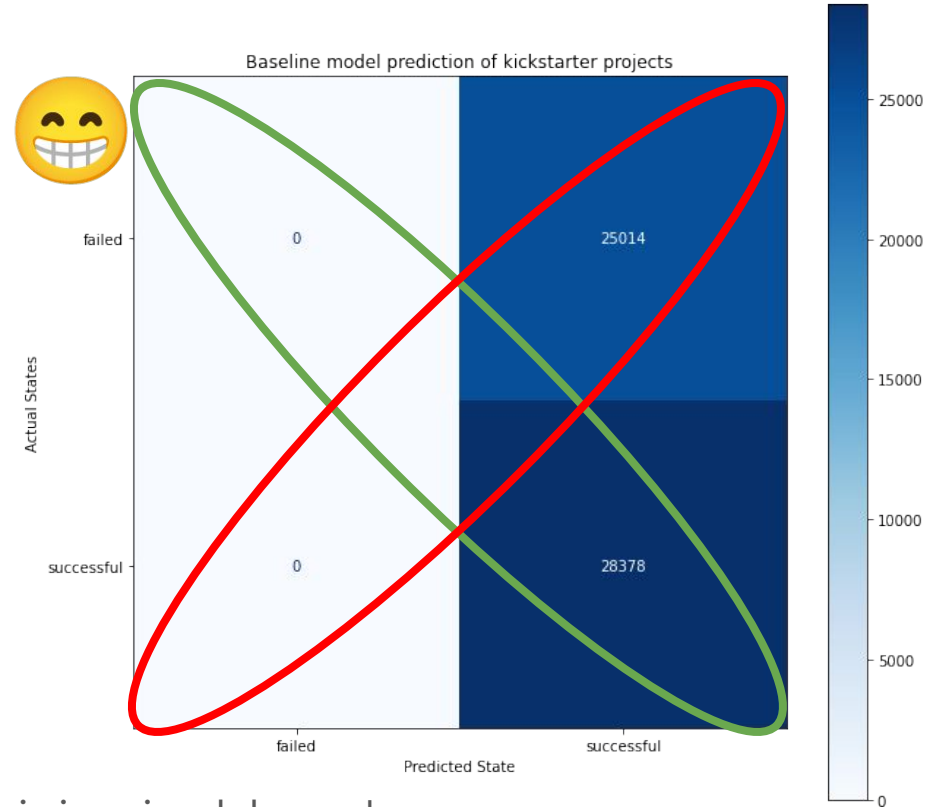
Baseline Model

- DummyClassifier is an extremely simple model
- projects are classified based on what the most common classification was in the dataset
- the DummyClassifier classifies all projects as successful, since most projects were successful
- Correct predictions: 43%



Confusion matrix, simply explained

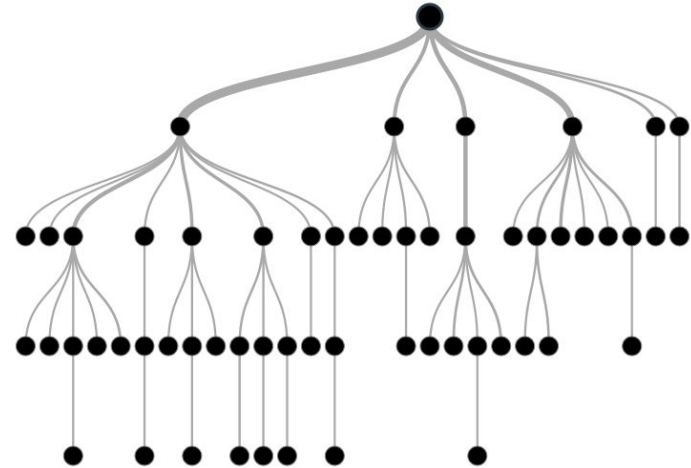
- Correct predictions
- Mispredictions



We want to minimize missed chances!
So we score our model accordingly

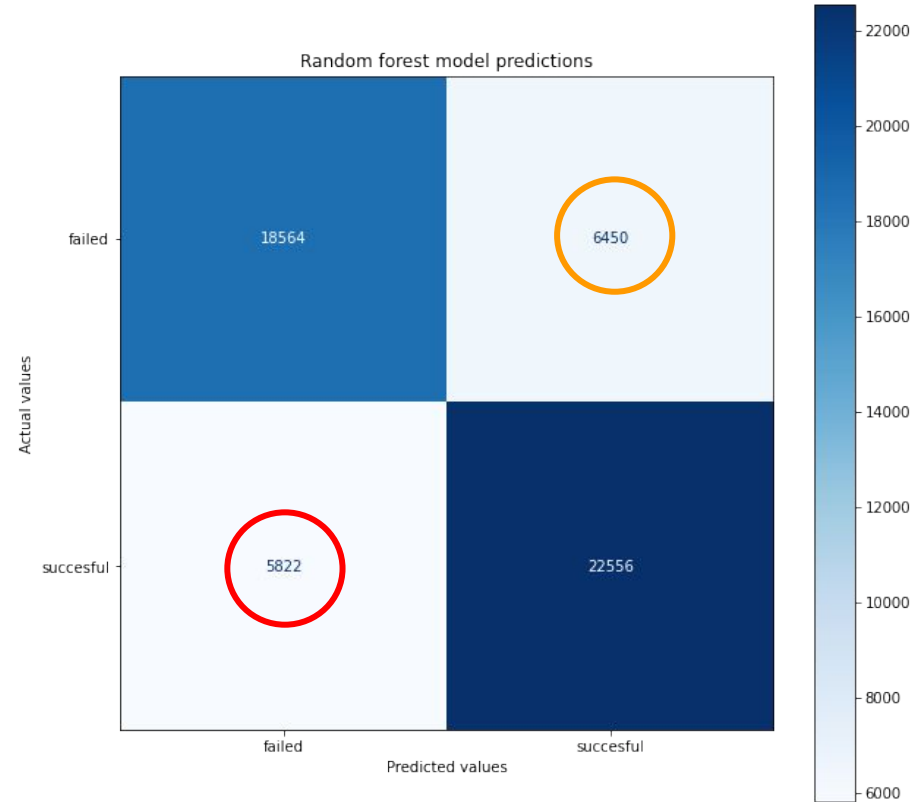
The model selection

- after cleaning and restructuring of the data
 - we had 406 features to model
- a decision tree model was chosen (random forest)
 - since we had a lot of categorical data
 - interpretability is important for goal-based consulting
- be able to tell our stakeholders
 - not only **if** projects have a high chance of being successful
 - but also **why** → **feature importance!**



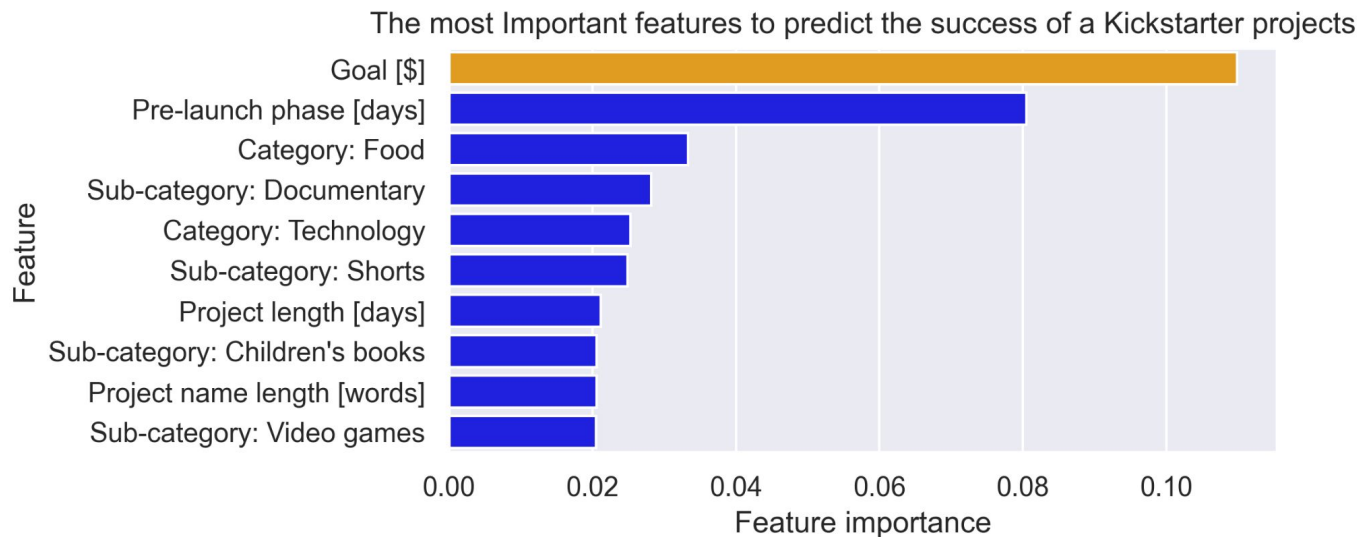
Random Forest Model

- Correct predictions: 77%
 - 79% increase compared to our baseline model!
- Fewer **missed opportunities** compared to **wrongly predicted successful projects**
→ desired outcome ✓

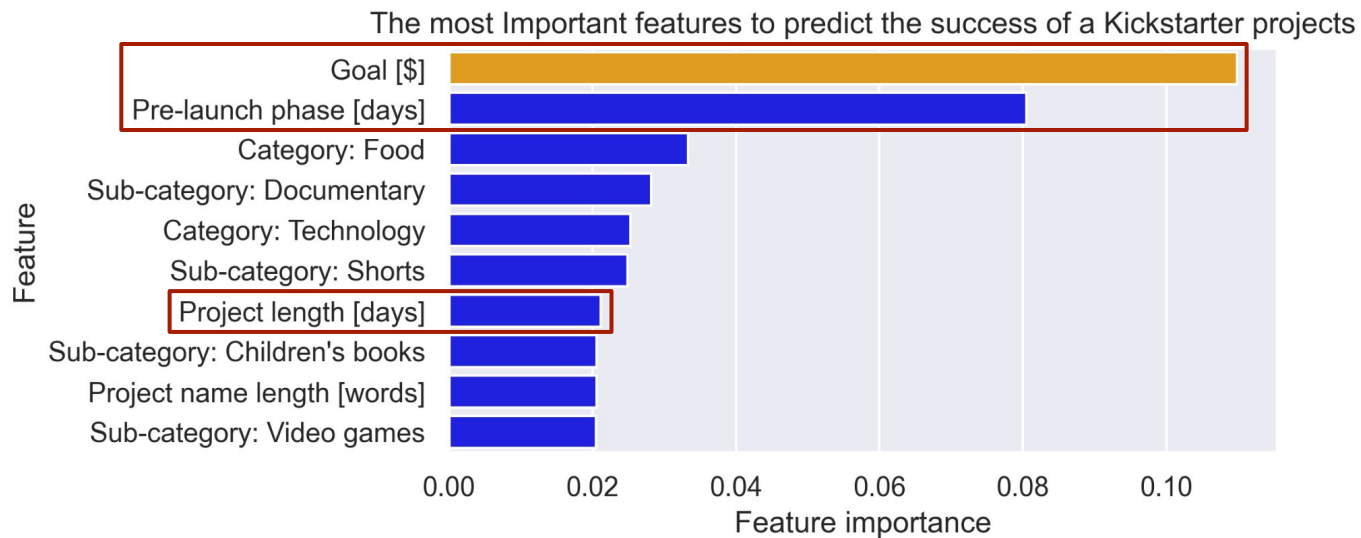




The most important factors

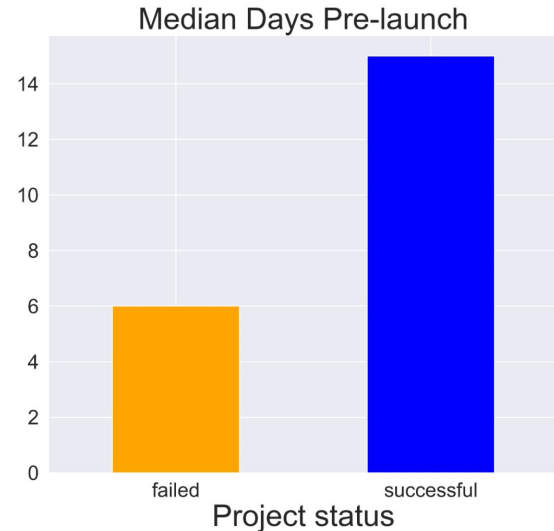


The most important factors

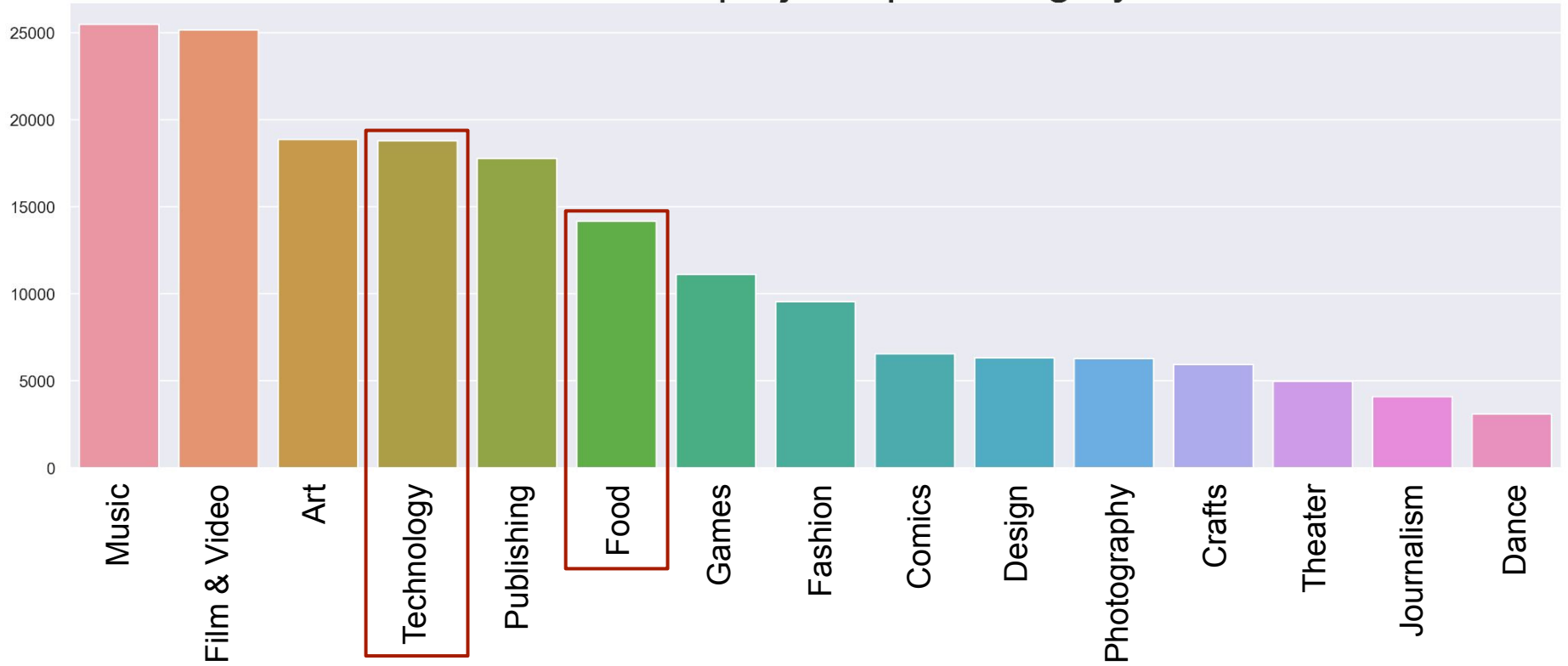


A closer look at the pre-launch phase

- the pre-launch phase is
 - the phase between project creation and the launch of the project
- this phase seems to be critical
 - a longer phase leads to more successful projects
- Kickstarter can support creators by
 - social media marketing
 - promotion videos
 - influencer marketing



Number of projects per category



The recommendations

- the funding goal is a crucial factor
 - lower funding goals lead more often to successful projects
 - is probably more realistic for the backers
- the pre-launch phase is very important
 - longer phases lead to more successful projects
- projects in certain categories are more successful
 - for example food and technology
- the project length has an influence
 - but it is only marginal



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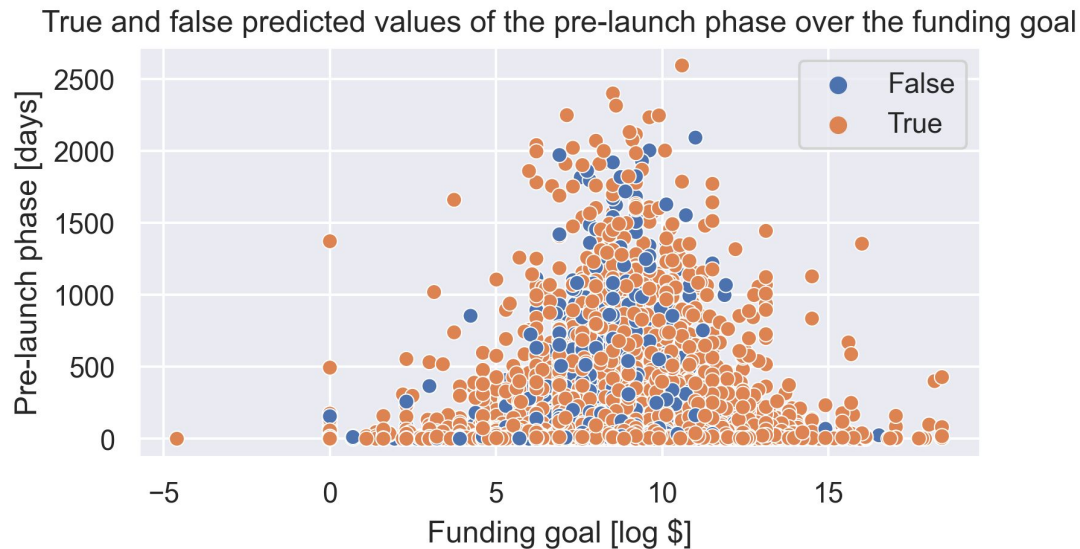
Outlook

- another model option would be the XGBoost (gradient boosting)
 - performs well for large number of (maybe irrelevant) features
- look at the different ranges of funding goals
 - there are a lot of projects with low funding goals (under \$10 000)
 - but also some with much higher goals

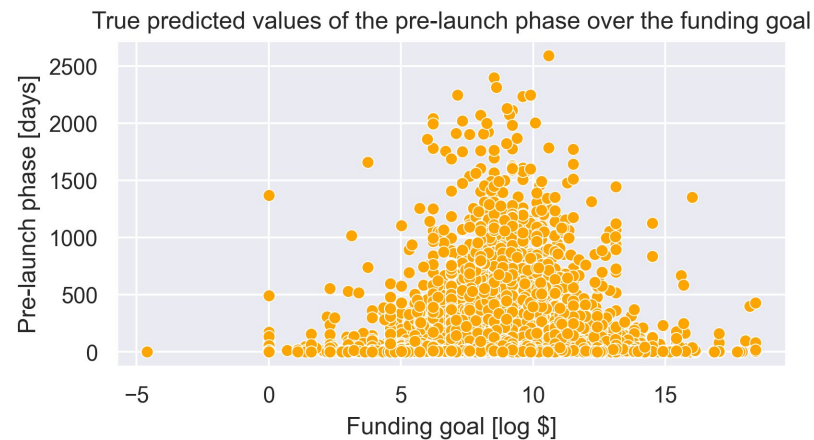
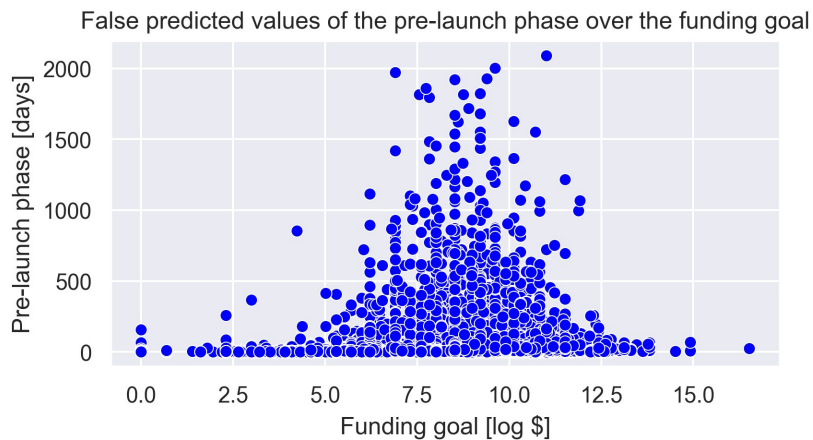


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Error analysis - True vs false predicted



Error analysis - True vs false predicted



Error analysis - Numerical variables

