#### HOW TO

## CROWDFUNDING A SUCCESS STORY

08th October 2020 Britta & Raphaela

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### What about kickstarter

- Kickstarter is a funding platform for creative projects - all-or-nothing model
- From films, games and music to art, design and technology
- As a creator you are responsible for setting your financial goal & fundraising deadline



#### What about kickstarter

Crowdfunding is an incredibly powerful tool to build communities and raise funds.

**60.97**% of all projects were successful and only **39.03**% failed.

#### What about kickstarter

If you have a great idea,
going on a platform
like Kickstarter can give you the
necessary push.

Most important features & label

# Data

O2 Data

- Kickstarter Data from 2009 2019
- **37** columns and **209.222** rows
- Features can be divided into different groups: projectinfos, amount / currency, datetime and campaign creation
- A project is successful if the pledged amount is bigger or equal your defined goal

Data

Füllfolie? tonsupur?

 A project is successful if the pledged amount is bigger or equal your defined goal.

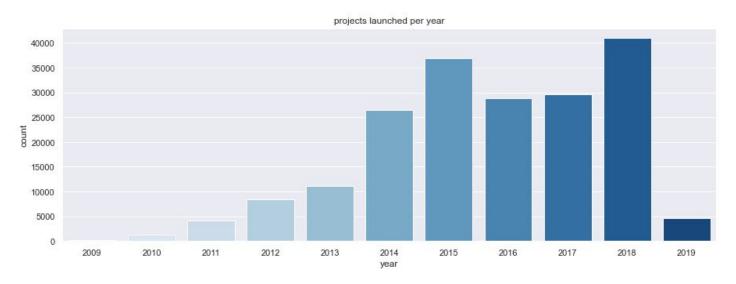
What can be done to make a project successful?

Success before project launch

# Be prepared

### Timing - year

• In 2018 in total 40.942 projects have been launched

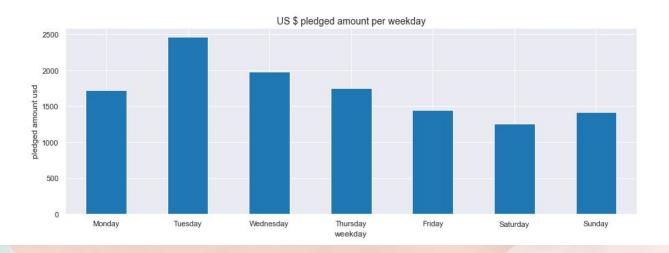


### Timing - month

- Most of the successful projects were created in the month
  - o January: 10.946
  - o March: 10.639
  - October: 10.407
- Most of the **projects failed** in the month
  - o December 4.672
  - February 5.935
  - November 6.004

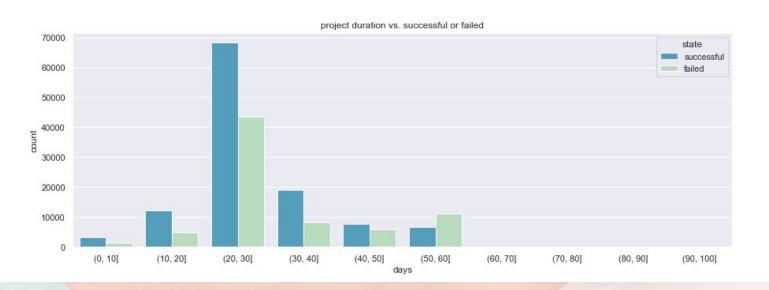
### Timing - weekday

- Most of the successful projects were created on tuesday (25.385), wednesday (20.836) or monday (20.719)
- Most money will be given on tuesday, wednesday or thursday



### Timing - project duration

- In average the project duration from launch to end is 30 days (median)
- Most of the successful projects had a duration between 20 and 30 days

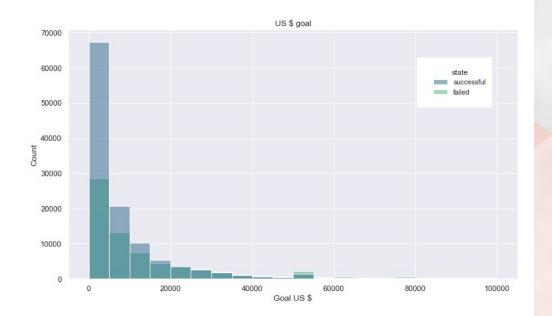


It's all about

# Money

### Money

 Don't be greedy - most of the successful projects do have a goal up to 20.000 US \$

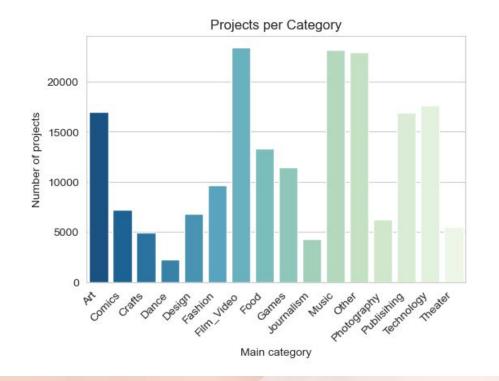


It's all about

# The Idea

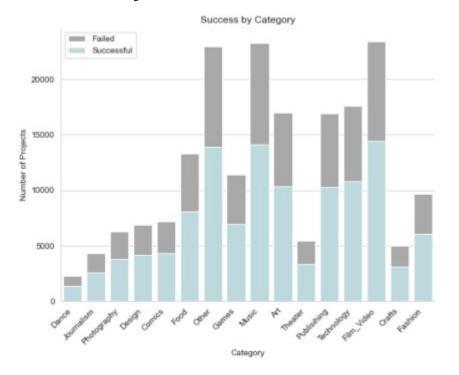
### Categories - Distribution

- There are **16** categories
- Some of the categories create stronger interests like film & video or music
- Not a big difference of success rate for each category



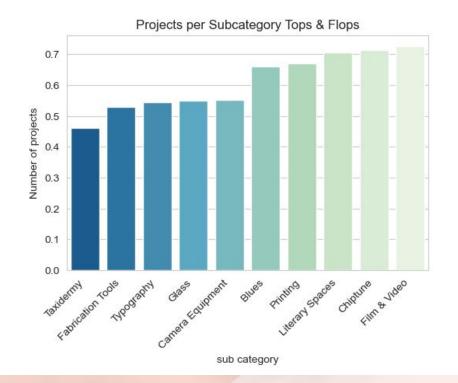
### Categories - sorted by success rate

The most successful category is Fashion.



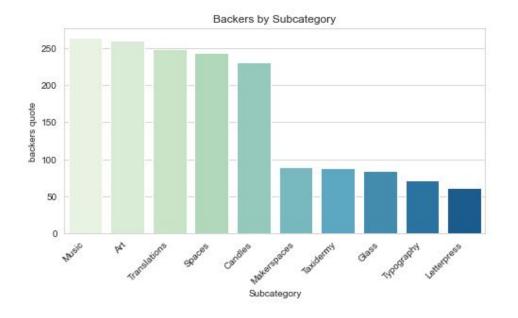
### Subcategories Top & Flop - Success rate

- There are almost 160 subcategories
- The success rate varies
   a lot from 46 %
   (taxidemy) to 72 %
   (film&video)



## Backersquote Subcategories **TOPS & Flops!**

- The number of backers per project depends on the subcategory!
- The hightest values
   exists for Music and Art
   with more than 250
   Backers per Project, the
   lowest value is 60
   Backers (Letterpress).

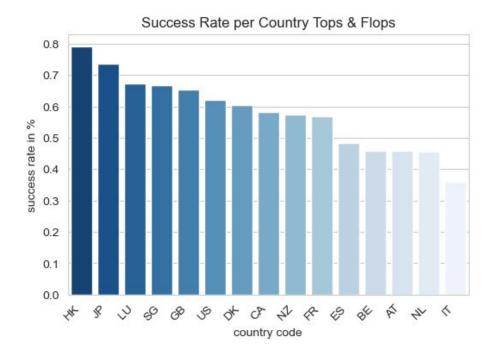


It's all about

# The Country

- The success rate varies between almost 80 % (Hongkong) and 36 % (Italy)
- Nevertheless 72 % of all projects come from the
   US

### Country

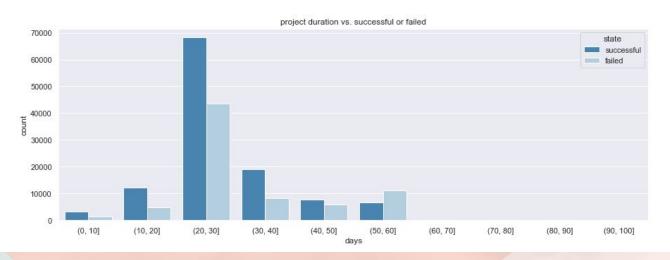


It's all about

# Being on point

### Being on point

- The **length** of your **projectinfo text** is important
- In average the projectinfo does have 124 letters / symbols (median)
- Most of the successful projects do have a length of 120-140 letters / symbols

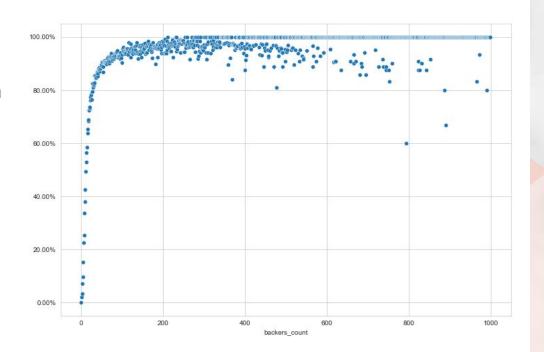


Success factors during runtime

# During runtime

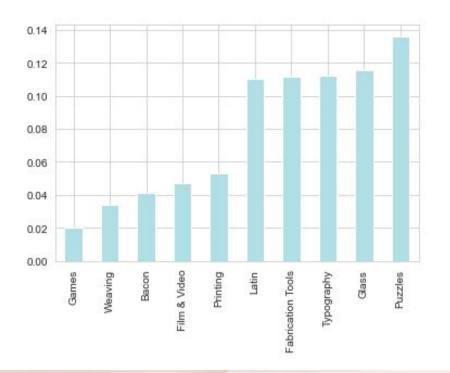
### More **Backers** are better!

- The success rate of a project highly depends on the number of backers
- Not very surprising, but what can be done to increase the number of backers?



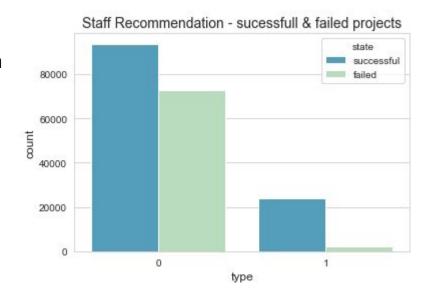
### Zerofollowers

- There are some projects without any backers
- The quote also depends highly on the subcategory
- The highest quote exists for **puzzles** (13,6 %), the lowest for **games** (2 %).



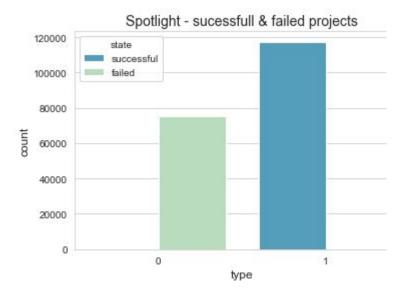
### Staff Recommendation

- The fact if a project is recommended by the staff has a hugh impact to the success of a project
- Only 9 % of the projects with staff recommendation failed compared to 44 % without



### Spotlight

- Spotlight on if a project is highlighted there is a
- 100 % chance of being successful



## To sum it up

Kickstarter campaigns looking to be successful should have a low goal, be film / video related, have a short duration be launched on a tuesday in january the right text length.

A good signal during runtime is having loads of backers, recommendations and a spotlight.

Success or failed

# Prediction

### **Prediction**

- We've created different models to predict if your project will be successful or not
- With a accuracy of 70 % and a precision of 72% we can tell you if you are going to be successful or not

What to come?

## Future work

## **06** Future Work

- Check the city
- Check the length and typo of the projectinfo text (blurb), slug and project name
- Check **medium**
- **Hyperparameter** tuning

# THANK YOU

Do you have any questions?

# Appendix

### Basismodel

After features selection we've choosen the following features for both models:

• main\_Fashion, month\_launched, static\_usd\_rate, day\_launched\_cat, year\_launched, currency\_cat, usd\_type, staff\_pick, goal\_usd\_log, backers\_count

From beginning the **logistic regression model** had very good accuracy and f1-score of. After hyperparameter tuning, the model's accuracy score could be improved slightly up to 0.90531 and the f1-score could be improved up to 0.92311. Also the precision got better from 0.991286 up to 0.91298.

The **random forest model** also had a an even better accuracy and f1-score. After hyperparameter tuning, the model's accuracy score got wores to 0.93943 and the f1-score got worse as well to 0.95096. Maybe we need to check that. But still very good values.

In this case we used for example the feature *backers\_count* and *staff\_pick*. In case of the very good results we came up with the result that both of these features can't be known before launching a project. Obviously an easy way to know that we have data leakage is if we are achieving performance that seems a little too good to be true. So we decided to make a new model in the following notebook.

### Model

From beginning all models had a quiet good accuracy and a f1-score. After hyperparameter tuning, mostly all the model's accuracy and f1-score could be improved slighty.

#### AdaBoost

#### Before:

Accuracy Score : 0.68868
 Precision Score : 0.70569
 Recall Score : 0.84087
 F1 Score : 0.76737

#### After

Accuracy Score: 0.69178
 Precision Score: 0.70675
 Recall Score: 0.84652
 F1 Score: 0.77034

#### **XGBoost**

#### Before:

Accuracy Score : 0.69827
 Precision Score : 0.70991
 Recall Score : 0.85547
 F1 Score : 0.77592

#### After

Accuracy Score : 0.70627
 Precision Score : 0.72649
 Recall Score : 0.83237
 F1 Score : 0.7758

#### **Random Forest**

#### Before:

Accuracy Score : 0.66067
 Precision Score : 0.67825
 Recall Score : 0.74802
 F1 Score : 0.71142

#### After

Accuracy Score: 0.664408
 Precision Score: 0.66777
 Recall Score: 0.795746
 F1 Score: 0.72616

### Slug length

- Most of the slugs are around 50 letters/symbols
- Which leads also to the most successful projects

