

Can you kick it?

Predicting what makes a Kickstarter campaign successful

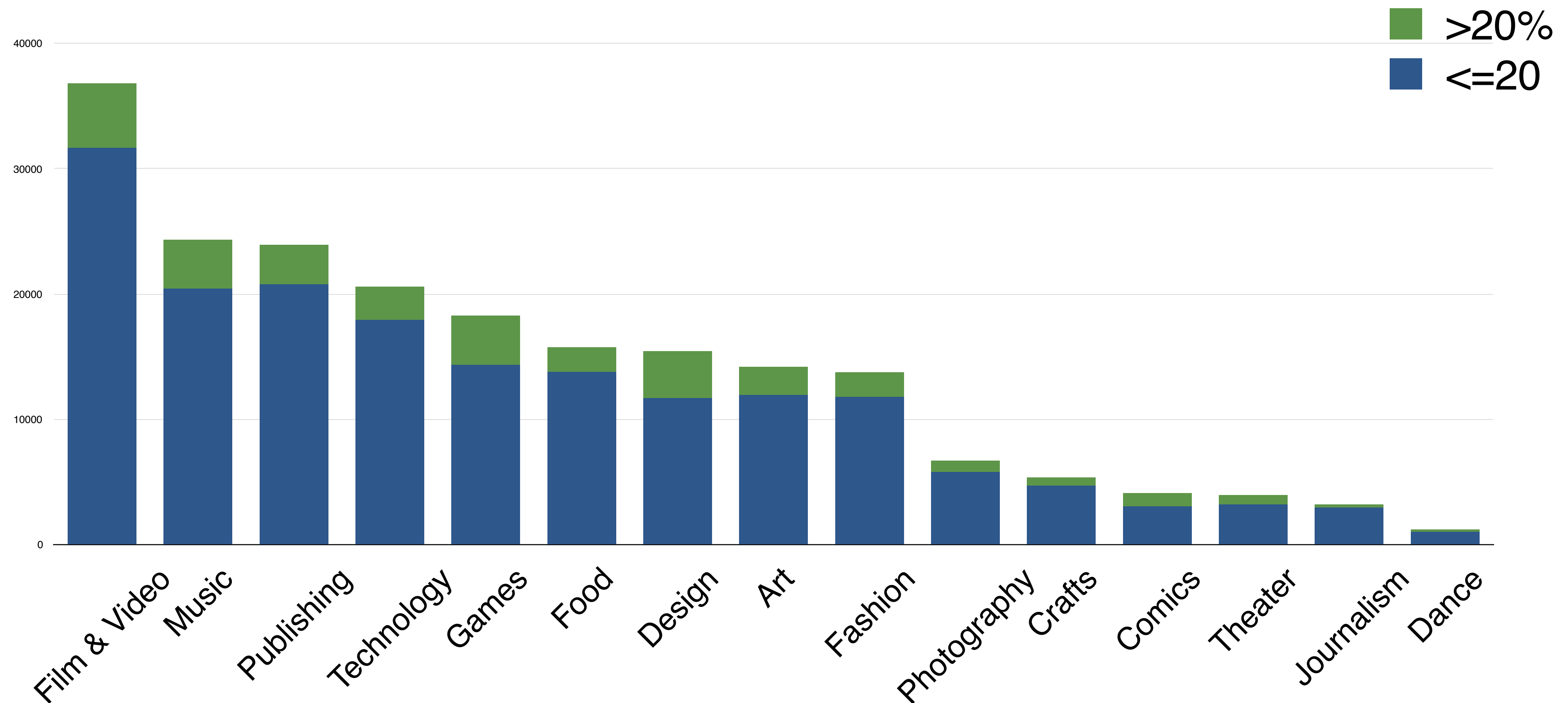
Elizabeth Sia

lizsia@gmail.com | <https://www.linkedin.com/in/elizabeth-sia> | <https://github.com/lizbug/>

Only 35% of Kickstarter Projects are Successful

and when they fail, they fail BAD

80% of unsuccessful projects **DO NOT EVEN REACH 20%** of funding goal



Solving this problem began with scraping Kickstarter campaign data

founder
updates/
comments

Campaign

Updates 15

Comments 71

Community

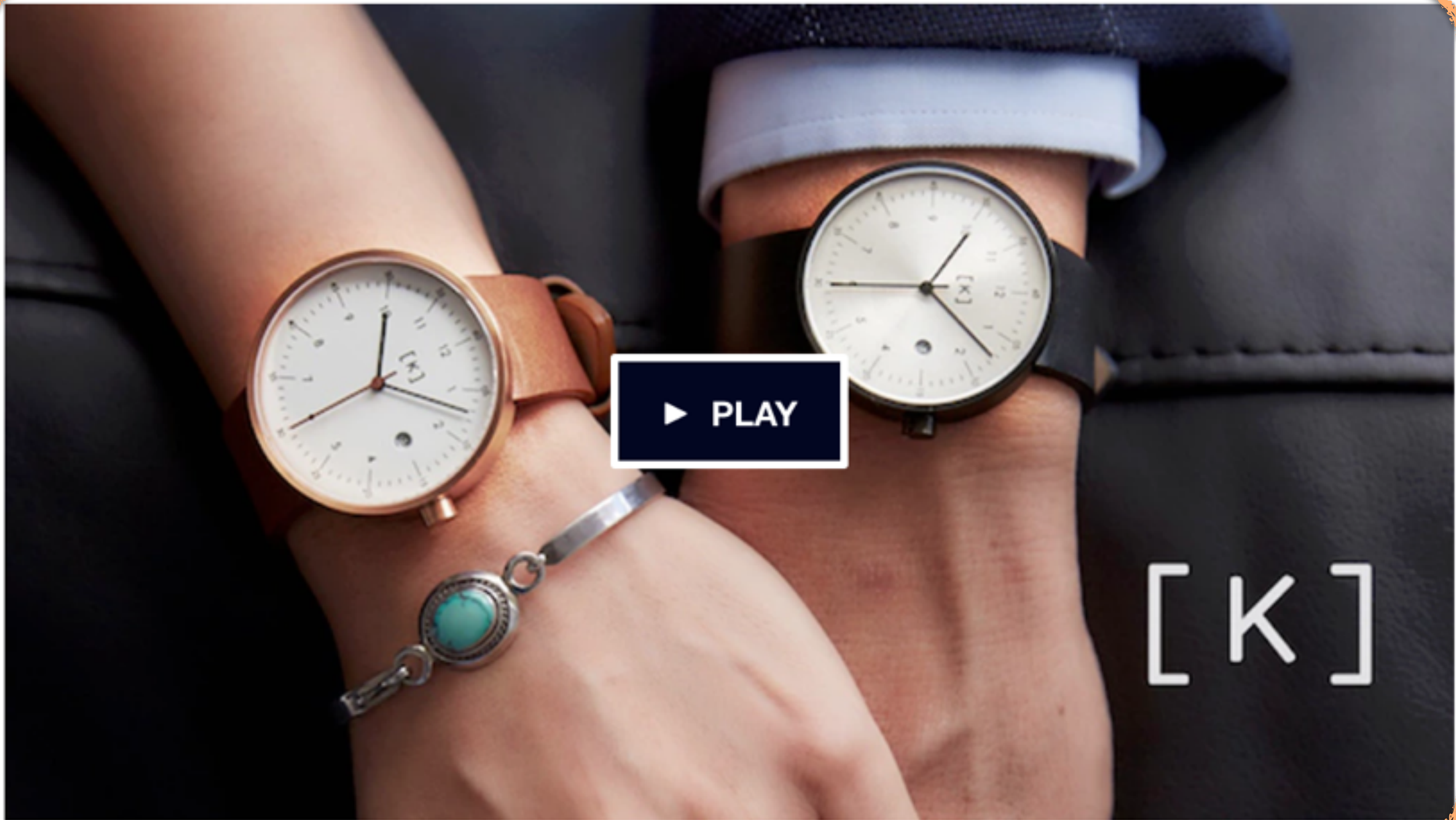
Share this project



About this project

Image/video content

Supporter Project reward structure



iKi 'A' Series Watch: Minimalist watch with Swiss movement

Melbourne, AU Product Design

AU\$ 23,622

pledged of AU\$ 15,000 goal

173

backers

Pledge AU\$ 1 or more

A big thank you from the iKi Studio for your support!

ESTIMATED DELIVERY

Sep 2016

6 backers

Pledge AU\$ 99 or more

Approximate 75 US/ 66 EUR
EARLY BIRD SPECIAL - 1 x 'A' Series iKi watch of your choice:

A01 - Rosegold casing with tan brown leather strap; OR
A02 - Gunmetal casing with black leather strap

50% off RRP of A\$199

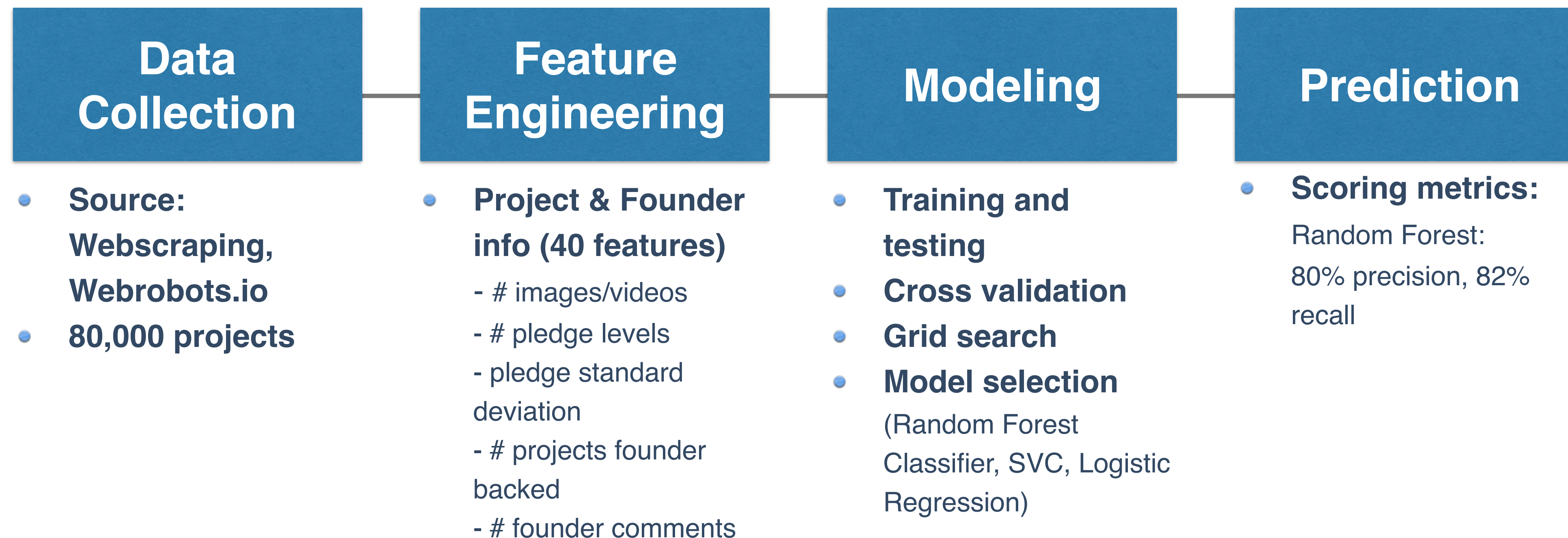
ESTIMATED DELIVERY SHIPS TO

Sep 2016

Anywhere in the world

Project title,
funding goal,
backers

Metadata-based classifier model yields 80% precision



The features that identified a successful project did not tell much about the project's story

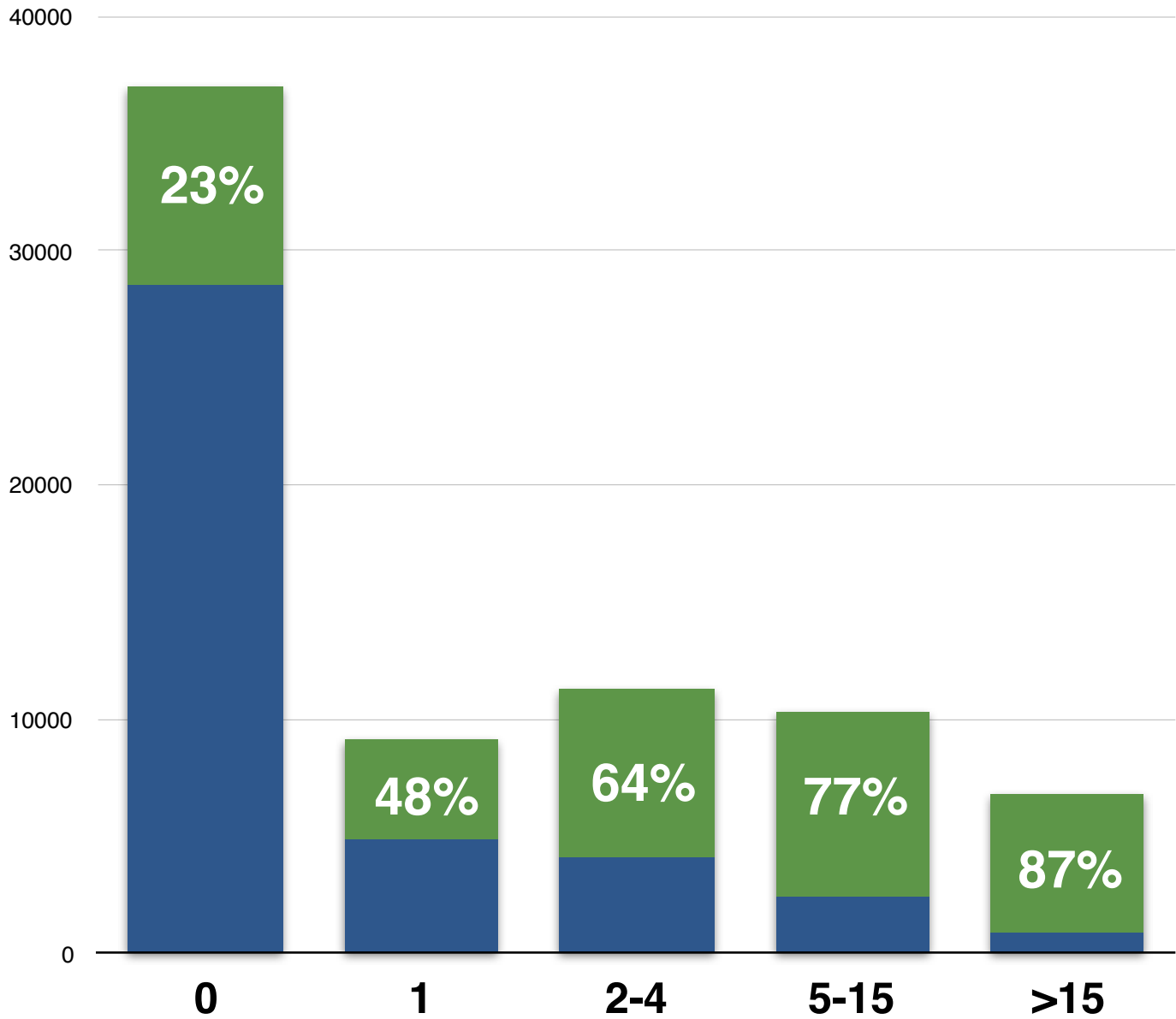
Successful founders backed more projects

Lower funding goals were likely to be funded

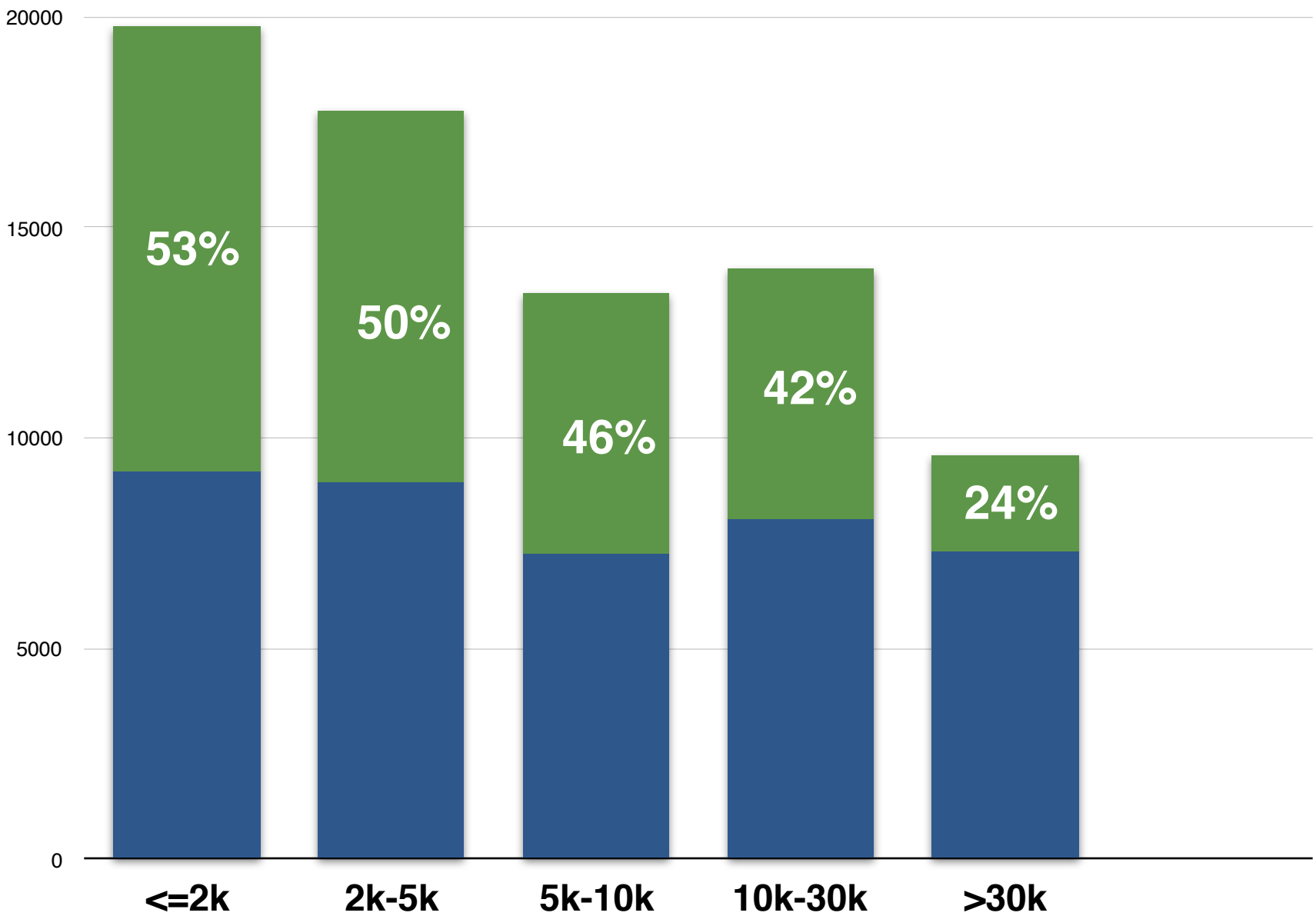
Successful projects have reward structures that require less # backers

■ Successful
■ Failed

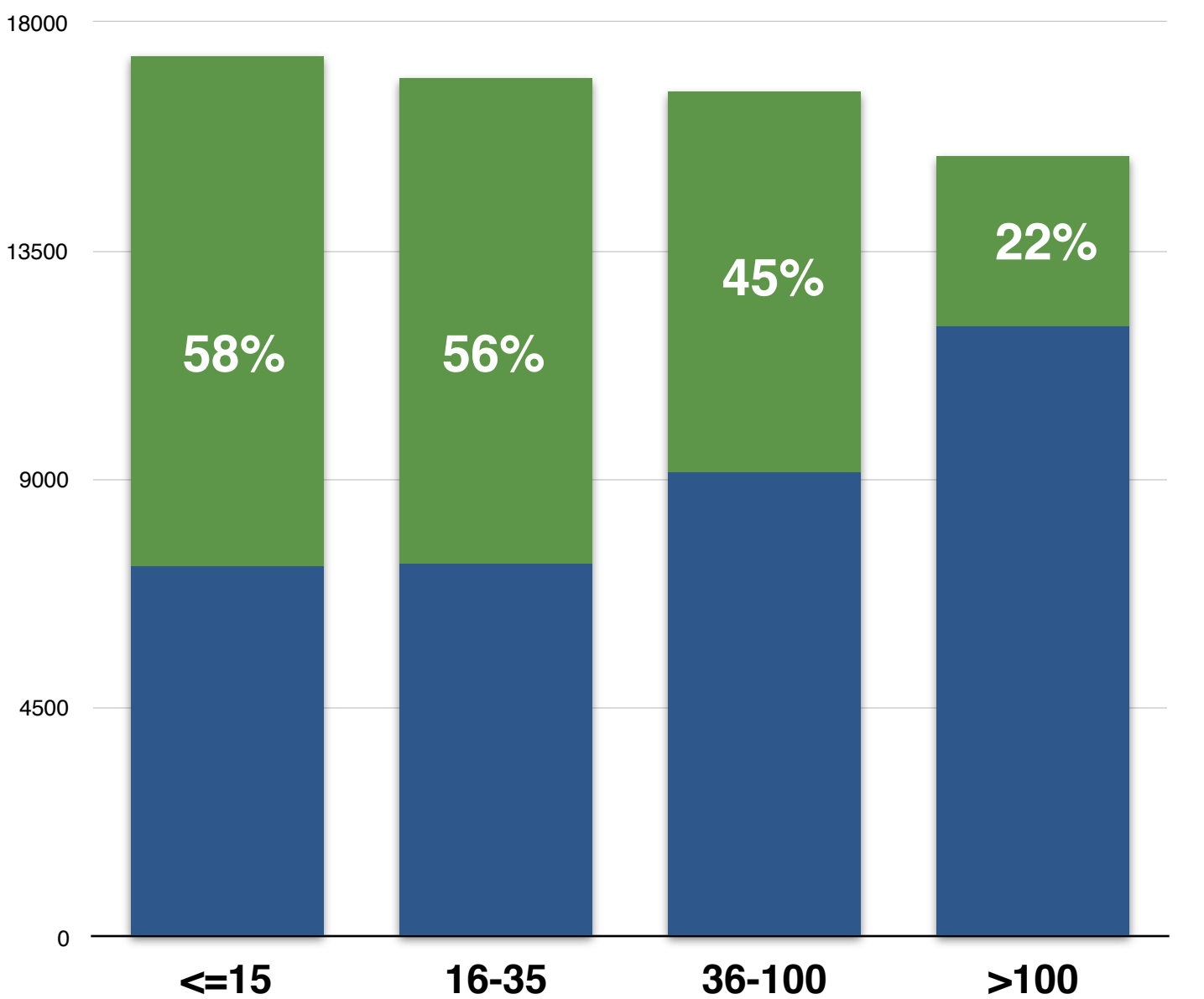
Breakdown by # of projects a founder backed



Breakdown by project funding goal

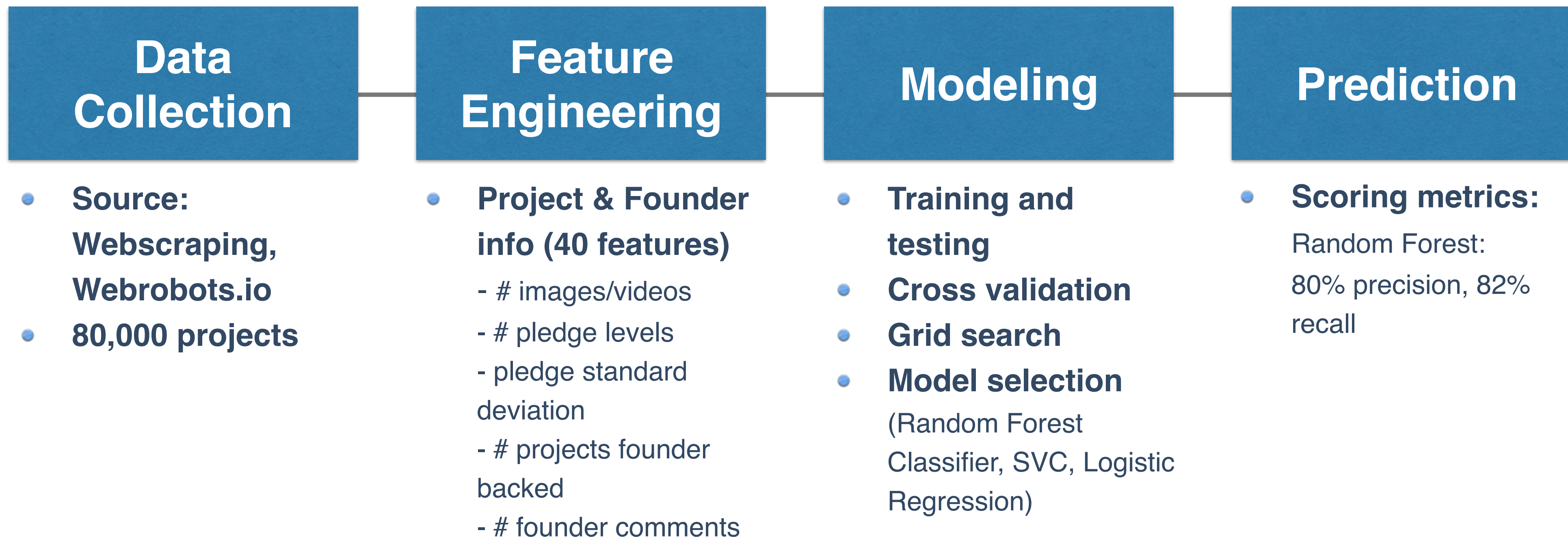


Breakdown by average number of backers required



Will adding text information from project descriptions build a stronger and more interpretable model?

Metadata



Will adding text information from project descriptions build a stronger and more interpretable model?

Metadata



Project descriptions

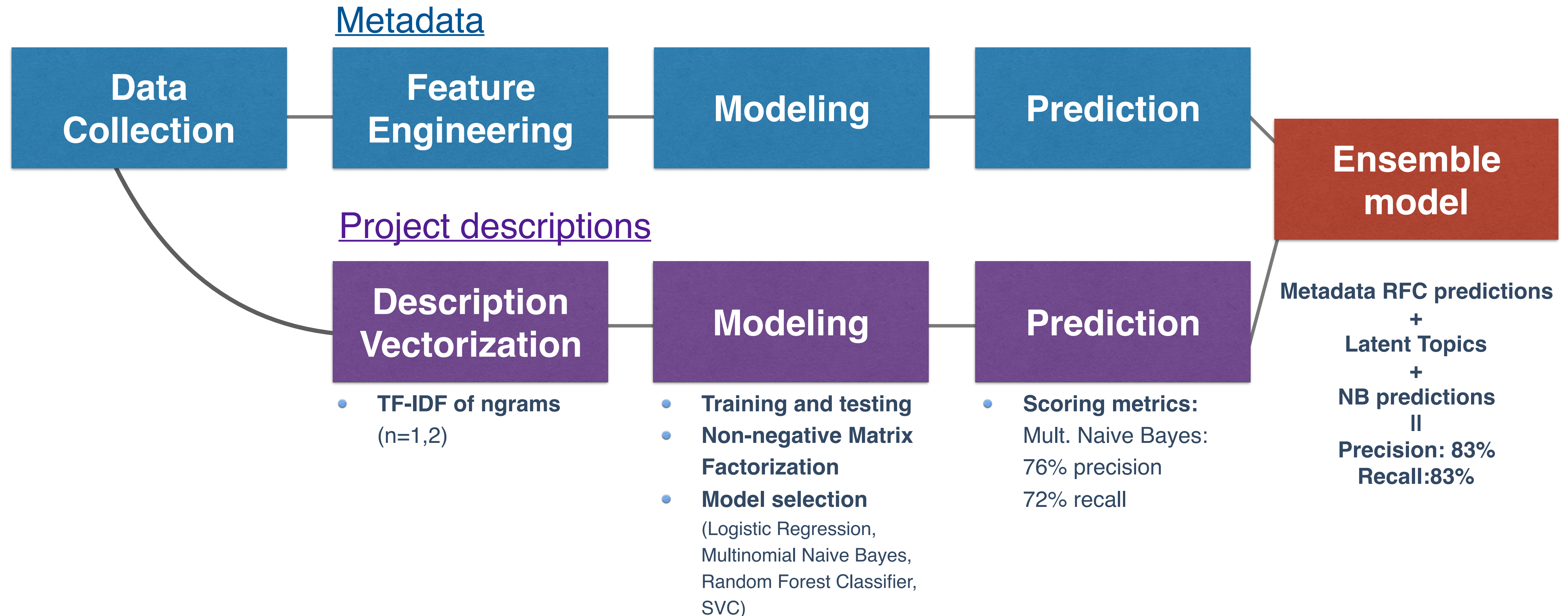


- **TF-IDF of ngrams**
(n=1,2)

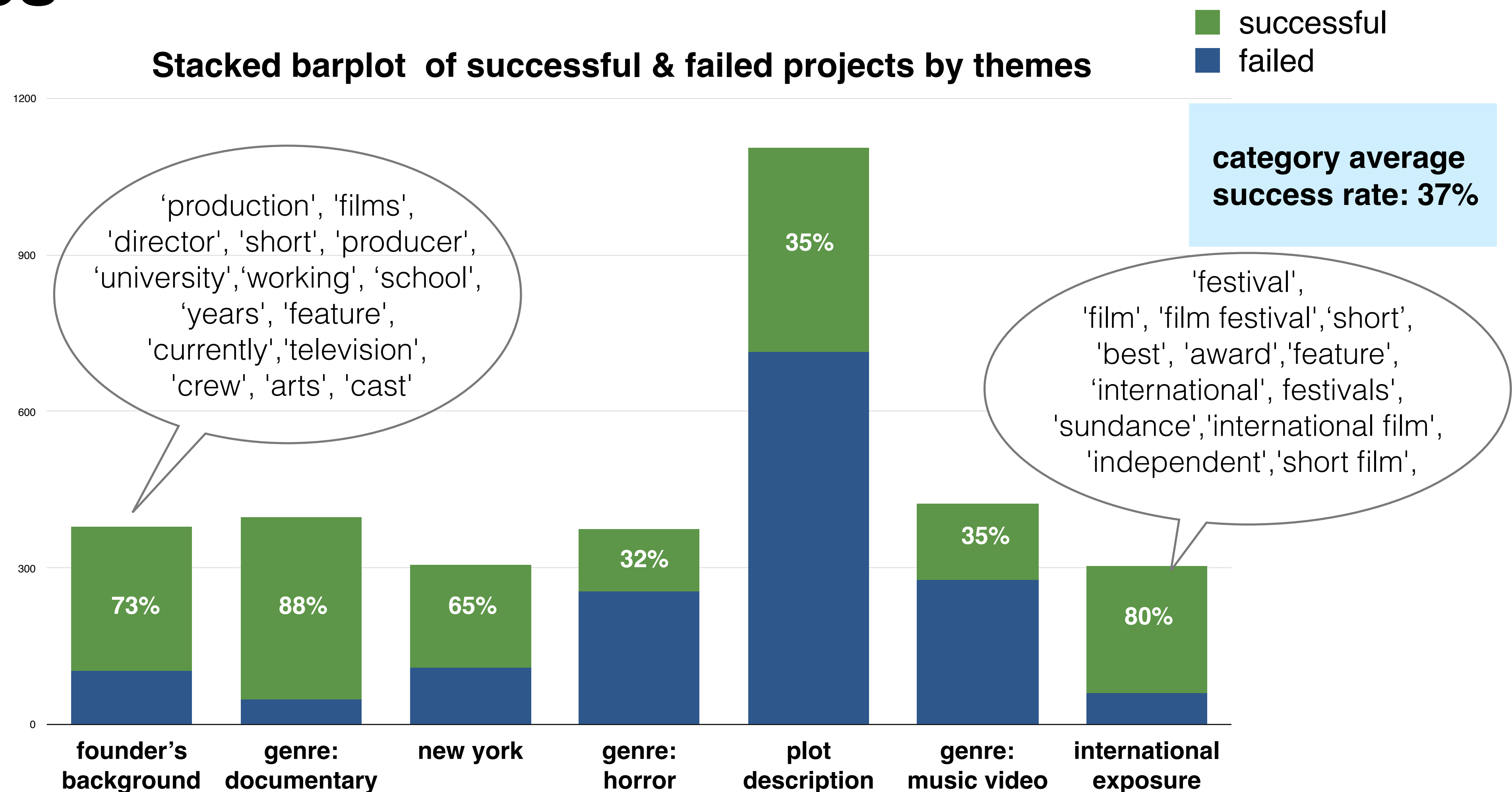
- **Training and testing**
- **Non-negative Matrix Factorization**
- **Model selection**
(Logistic Regression, Multinomial Naive Bayes, Random Forest Classifier, SVC)

- **Scoring metrics:**
Mult. Naive Bayes:
76% precision
72% recall

Ensemble model captures metadata and latent topics signals to increase score to 83%



Successful projects followed certain underlying themes



Future work

- Optical character recognition to detect text within the images
- Explore additional features (sentiment analysis of comments and temporal features)

Thank you

Elizabeth Sia

lizsia@gmail.com

<https://www.linkedin.com/in/elizabeth-sia>

https://github.com/lizbug/Capstone_Project_CanYouKickIt

Appendix

Feature importances from metadata Random Forest classifier

