# KICKSTARTER

Supporting Kickstarter management to help creators succeed on the platform to drive platform growth

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# **Executive Summary**

Introduction	<ul> <li>Introduction to Kickstarter's business and slowing growth trend in recent years</li> <li>Kickstarter can do more to help creators increase project success rates to improve our platform uptake and growth</li> </ul>
Objective	<ul> <li>Identifying factors that affect the success of projects to guide creators in marketing and producing better projects that will increase repeat backers and position Kickstarter as a go-to platform for creators with higher chances of success</li> </ul>
Dataset	<ul> <li>The dataset consists of projects from 2021 to 2022</li> <li>Contains the details of each project such as the goal, amount raised, deadline and the state of project</li> </ul>
EDA	<ul> <li>Understanding success rates by category and sub-categories</li> <li>Analysing goal settings and pledge amounts over time as an indicator of project success</li> </ul>
Model Development	<ul> <li>Exploratory data analysis coupled with objective assessment to identify potential relationships such as multicollinearity between variables</li> <li>Oversampling via model weights to account for data imbalance of significantly higher number of successful projects in our dat aset</li> <li>Ran random forest, Logistic regression and XGBoost</li> </ul>
Model Evaluation	<ul> <li>XGBoost chosen as the preferred model with ROC of 0.919, accuracy of 0.842</li> <li>XGBoost yields low false positive rate of 4.9%</li> <li>Goal, Deadline Day (Sat/Mon), Main/sub-category, Launch Month, Launch Day (Sat), Launch Duration, Blurb Sentiments, Project Duration found to be key features</li> </ul>
Limitation	<ul> <li>Previously successful projects are not accounted for as a handful of Kickstarter projects are a continuation of previous projects and many projects are dependent on the success of previous projects due to the pre-existing large pool of trusted backers</li> <li>Project features relating to level of engagement unavailable (e.g. number of updates or responses to comments) reflecting level of engagement of creators with backers which could be another key factor for determining a project's success</li> </ul>
Recommendation	<ul> <li>KickOff – Interactive Tableau dashboard for creators to understand the success rate and funding target of similar projects / projects in the same category and/or subcategory across different geographies</li> <li>KickFar – Campaign success prediction dashboard on Heroku allows creators to predict project success</li> <li>KickHigh – Helps struggling creators succeed by partnering with equity crowdfunding platforms</li> </ul>

# Kickstarter's AON model passes on risks to creators but more can be done to incentivise creators to succeed on the platform



## **Understanding our datasets**

Main Projects Dataset

Information: 34269 rows x 39 variables Retrieved from: Web Robots - <u>https://webrobots.io/kickstarter-datasets/</u> Retrieved on: April 16, 2022

Key Variable	Description					
Backer Count	Total number of backers that have backed the project					
Blurb	Project Description					
Category	Main Category and Sub-Category					
Converted Pledge Amount	Total pledged amount converted to the default currency of the project from multiple currencies (if any)					
Country (Displayable Name)	Country Code and Country Name where the project is in					
Created At	UNIX timestamp of the Project Creation					
Creator	Details of the Project Creator					
Currency (Symbol)	Default Backing Currency of Project					
Deadline	UNIX timestamp of the Project Deadline					
FX_Rate	Standardised USD:Currency FX Rate					
Launched At	UNIX timestamp of the Project Launch					
Location	Location details (Lat Long, Country etc.)					
(USD) Pledged	Pledge amount in chosen currency and USD					
Spotlight	Kickstarter show cases successful projects					
Staff Pick	Kickstarter staff choose which projects they like					
Problem Analysis	Exploratory Data Analysis Model Develo					

### **Time Series Dataset**

Information: 600000 rows x 10 variables Retrieved from: http://sidekick.epfl.ch/data<sup>1</sup> Retrieved on: April 22, 2022 Additional Pre-Processing done to extract from NumPy array files

Key Variable	Description				
ld	Project id, used as unique identifier				
time_interval	Time of the sample (betw een 0 and 1, relative to the start and end of the campaign				
Pledge_amt	Current amount of pledged money (relative to the campaign's goal, so multiply with the goal to obtain the real amount)				
No_backers	Current number of backers (usually not an integer, because the statuses have been resampled to have an equal number of statuses)				
Goal	Project goal state: Project final state (1=successfully funded, 0=failed)				
Launch_date	Launch date (as a UNIX timestamp)				
Deadline	Deadline (as a UNIX timestamp)				
Duration	Duration of project in days				
actualpledge	Actual pledged amount, already multiplied by campaign goal				

Source: 1. (Etter, Grossglauer & Thiran, 2013)

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## **Objective:** To help creators succeed on Kickstarter and drive the flywheel of growth

### Main Objective: Help Creators Succeed on Kickstarter

Kickstarter Fees Fixed: 5% of total funds raised Payment Processing: 3 - 5% + S\$0.30 per pledge (varies by country Total: ~8 – 10% of funds raised

Management's Priority Kickstarteris incentivised to help creators succeed on the platform as it drives revenue for the company

More creative projects instils confidence and appeals to a wider range of Kickstarter backers 1. P 000

More repeat backers position Kickstarter as a go-to platform for creators with higher chances of success Helping creators first will drive the flywheel

of growth for Kickstarter

### Data proves that success drives future confidence



Link to tableau analysis: https://public.tableau.com/app/profile/adelvn.koh/viz/ConsultingforAnalytics-Kickstarter2/ProjectOwnerDashboard?publish=yes

#### Where do the backers' investments Select Overall Category for Deep Dive oduct design qo? design \* .01M NΔ \$12.41M This dendogram provides an tovs interactive view of backers \$6.73M nteractive design: \$2.72M design breakdown, while the right side show raphic design: \$139.57M \$1.98M technology architecture: \$74.01M sub-category has received \$0.49M food civic design: \$63.25M \$0.18M publishing typography: \$62.19M \$0.05M comics Overall Investment \$51.57M film & video Breakdown \$46.52M music \$999,008,682.00 \$27.93M iournalism \$21.77M theater \$8.25M art \$2.57M crafts \$1.89M

Technology is one of the categories with lowest success rates but the second highest investment flow, which suggests backers are idea-Management can partner with an gel investor communities / venture capital funds to bolster seed investments in weaker categories

**Exploratory Data Analysis** 

## Identify Platform Fit: Amount Pledged by Category

## EDA – Understanding project success rates and making them visible to creators

By Category **By Sub-Category** Success Rates by Subcategory 50.00% 0.00% Highest Success rates vary greatly within sub-Comics, Design Publishing categories, which suggests that there Technology projects have the *highest* success Insiahts taraets, which makes it difficult for Insights are multiple factors involved – fundina Lowest Craft, Technology them to reach goals taraets, backer demand and (Only 1 in 3 succeed) *competition* success The importance of goal setting Goal Range for Similar Projects (# of projects with that goal range) Avg Goal Range **Goal Rationalisation** 14.322 11.912 8,158 6.284 5,316 5.020 \$60,741 3.360 2,424 2.030 1,962 1 352 746 790 886 190 0 1000 3000 4000 6000 8000 9000 10000 11000 12000 13260+ On average, successful Aggregated Amount Raised (Binned) projects had much \$8.241 lower (~7x) goals than Majority of successful projects (60.4%) set goals that are < 5000 USD. failed projects Cultivating the This suggests that herding and certainty economic factors failed successful influences backers to fund projects that are more likely to meet perception of success their goals. Source: Kickstarter, Lit Analysis, Team's tableau dashboard **Exploratory Data Analysis** 

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Success rates can be influenced by multiple factors, but funding target is one major factor

## EDA – Understanding project success rates and making them visible to creators



### **Pledged amounts over funding duration**

#### Our goal is to make it easily visible to them what targets made past projects successful

Success rates (%) and No. of Projects over time



#### **Guidance is important** to creators

Number of projects picked up in 2014 and success rates grew steadily over time since then. This coincides with the release of the *creator's* handbook in 2014, which suggests that creators constantly seek and adopt advice to make their fundraising successful

#### **Kickoff – Project Owner Dashboard**

Simple, interactive dashboard for potential creators to understand the success rate and funding target of similar projects / projects in the same category and/or subcategory across different geographies

#### Whv?

While creators should think about funding targets based on their own needs, this dashboard helps in goal-rationalisation and budgeting.

Link to Kickoff: https://public.tableau.com/app/profile/adelvn.koh/viz/ConsultinaforAnalytics-Kickstarter2/ProjectOwnerDashboard

## **Model Development**

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#### We will thus be using the following features for our models

Goal

launch vea

created year

deadline year

blurb length

name length

blurb\_sentiments

- Parent category
- Child category
- Campaign Duration
- Launch Duration
- Launch Month
- Created Month
- Deadline Month

- Created Year
- Launch Day
- Created Day
- Deadline Day
- Country
- Blurb Length
- Name Length
- Blurb Sentiment Score

Source: Dataset used excludes certain countries such as Russia due to lack of availability

#### Problem Analysis

Exploratory Data Analys

# Severe Collinearity

**Model Development** 

created year,

deadline year and

launch\_year

Some Collinearity

created month,

deadline month and

launch month

Decision

Projects can be listed up

to 60 days, thus, the month information is critical and will be kept.

## Scaling/ Standardize of Data

As we are utilising PCA analysis for dimensionality reduction, we need to scale the data to standardize columns prior to model fitting.

Scaling and Oversampling the data

## 2

Oversampling of Data (Assigning different weights instead of SMOTE



Thus, **oversampling is necessary** to ensure both failed and successful projects are equally represented in the model and will be less biased towards predicting successful projects. (we assigned different weights)

#### Model Evaluation

#### Conclusions

## Choosing and evaluating the best model – XGBoost

## XGBoost is chosen as the preferred model

Method	ROC-AUC	Precision	Recall
Logistic Regression	0.89	0.77	0.80
Random Forests	0.88	0.79	0.81
XGBoost	0.919	0.919	0.835

#### XGBoost outperforms random forest and logistic regression models

Method	Training Accuracy	Validation Accuracy
Random Forest	0.95	0.83
XGBoost	0.92	0.83

#### Random Forest suffers from slightly more severe overfitting



XGBoost emphasises on functional space over hyperparameters in model optimization which is more **efficient** 

 $\mathsf{XGBoost} \text{ is } \textbf{good for unbalanced dataset}$ 

#### XGBoost is less likely to overfit

### **XGBoost** is the model of choice

Metrics	Scores
Accuracy	0.842
Precision	0.919
Recall	0.835
ROC-AUC	0.919

The ROC plot shows a curve that is close to the top-left corner, indicating a relatively good performance.



False Positive Rate



## **Evaluating XGBoost on Test Set**

Problem Analysis

## Which features are the most important in project success?



#### Insights

#### Why do we think it's important (Intuition)



Goal Amount is the most important feature – makes intuitive sense as campaign success is decided by whether pledge amount meets the goal

2 Categor Preferer	ical <b>3</b>	Launch Timing, Day	/	4	Project Descriptions
Categories suc Games/ Design attract more ba than categories such as Crafts/ Journalism as t likely appeal to wider range of audiences	h as Pro , gei ckers ser gai mo hey lau a cer as	oduct releases ar nerally time nsitive. i.e., a vide me might attract ore backers if nched during tain seasons suc holidays	e o h	A wel might intere and le backe	I-crafted blurb drum up more st in a project ead to more er interest.

#### How can management act on these insights?

Educate creators on how they can utilize the dashboard by inputting variables such as launch dates to view the success rates of other projects with the same variables

**Model Evaluation** 

## KickFar Initiative – Allowing creators to predict campaign success

# We have designed a predictive dashboard with our model to provide actionable insights to creators...

# ... to predict the success of their campaigns before launch





#### Link to predictive model dashboard: https://predict-kickstarter.herokuapp.com/

Model Development

## KickHigh – the all-round solution to combat high targets and stretch goals



## Conclusions

	Our Winning Strategy		Limitations and Recommendations
		<b>1</b> Limitation	15
<b>KickOff</b> Dashboard for interested creators-to-be	Help creators-to-be decide whether Kickstarter is the right platform to raise funds and rationalise fundraising targets	Reiterated Projects	<ul> <li>A handful of Kickstarter projects are a continuation of previous projects, and are creating new and improved versions of a product</li> <li>These projects' success are highly dependent on the success of previous projects and already have a large pool of trusted backers</li> </ul>
KickFar Predicting	Help creators understand their	Level of Engage- ment	<ul> <li>Features such as the FAQs displayed, number of updates or responding to comments are unavailable and hard to collect</li> <li>Reflects the level of engagement by the creator which is crucial in building confidence and trust among the backers</li> </ul>
campaign success for	from campaign parameters	2 Model life	cycle and management
creators		Objective	<ul> <li>Capture new data for continuous learning</li> <li>Retrain models so they continually adapt to the dynamically changing customer segments of the company</li> </ul>
KickHigh Project creation	Help to direct creators who struggle to raise funds or whose	Benefits (Context- ualised)	<ul> <li>Prioritise projects that have high chance of success but might fail due to lack of visibility</li> <li>Track effectiveness of creator's success rate</li> </ul>
& goal guidance for creators	needs extend beyond small raises to equity crowdfunding platforms	Future consider- ations	<ul> <li>Adopt the use of alternative data – competitor data, credit card data (when signing up as a member), geospatial data to further segment the customers by income and location</li> </ul>

Problem Analysis

Exploratory Data Analysis

Model Development

Model Evaluation

## Appendix - More Initiatives – Improvements to Project Creation Dashboard

# We can further improve the predictive dashboard by actively recommending creators how to improve their campaigns..



## ... through our all-improved project creation dashboard



Problem Analysis

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Funding goal	Goal amount
Set an achievable goal that covers what you need to complete your project.	S\$ 0
Funding is all-or-nothing. If you don't meet your goal, you won't receive any money.	Use our calculator to estimate total costs, including taxes and fees.

#### Pledge Breakdown

The pledge breakdown at the top of the Fulfillment dashboard will help you to gain a clearer understanding of the final number of backers on your campaign page, and how this relates to fulfillment. You'll see the total pledges on the campaign, including the number of pledges with rewards and the number of pledges without a reward.

If your campaign had any dropped pledges, a count will appear alongside the other pledge numbers. If there are no dropped pledges however, this information will not be displayed.

#### Project description

Describe what you're raising funds to do, why you care about it, how you plan to make it happen, and who you are. Your description should tell backers everything they need to know. If possible, include images to show them what your project is all about and what rewards look like. Read more about telling your story

Headline	:=	в	i	e	L	 <b>()</b>

Write about your project like you're explaining it to a friend...

#### Allows individuals to look at more detailed analysis of the trained model

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Model Explainer	Positive class:	*				Feature Importances	Classification Stats	Individual Prediction	s What if	Feature Dependence	Decision Tr
Feature Importances	Classification Stats	Individual Predictions	What if	Feature Dependence	Decision Trees	Select Index	m		Prediction		
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Cutoff prediction probability:						Observed state:	X *	Range:	o label p	robability	3.3%
Cutoff percentile of samples:	0.25	c	50	0.75	о 0.99	Predicted probability range:	0.4 0.6	0.8	0 3.	.3 %	
0.01	0.25	C	50	0.75	0.99				* indicates observed	label	N6.7%
Model performance m	netrics		Confusion Matrix How many false positives and	false negatives?							
metric	5	Score				Contributions Plot How has each feature contribut	ted to the prediction?		Partial Deper How does the predic	ndence Plot tion change if you change one feature?	
accuracy	(	0.842		Confusion Matrix		Index:	Depth: Sorting:		Feature:	Index:	
precision	c	0.919	-0	.5-		2019 × -	High to Lo	0.5 774	goal	€ 2019 ×	
recall	0	0.835		- 29.2% 4.9%		100	bution to prediction probability		Shap Dependence		
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Problem A	Analysis	Explorato	ry Data Ana <u>lysi</u> s	s <u>M</u> o	del Develop <u>mer</u>	it	Model Evalua	tion	Co	nclusions	17

## **Appendix – Additional Analysis (Textual Analysis)**

#### Hypothesis: There are specific keywords that make a project listing more appealing / less appealing than others



#### Probability Density Curve of Projects by Blurb Length based on Category



#### Short and Sweet: Most projects have less than 35 words in their blurb

	word	success 🖕	success_rank \$	fail \$	failed_rank \$
1	album	1462	1	316	13
2	book	1143	2	454	6
3	world	959	3	504	4
4	music	884	4	524	3
5	story	811	5	240	25
6	series	803	6	297	15
7	love	731	7	372	9
8	graphic	682	8	28	590
9	life	679	9	449	7
10	film	664	10	262	20

We analysed unigram frequency of titles between successful and failed projects to notice any difference:

- 1. Most words appear in equivalent proportions for both successful and failed projects (on a ranking basis)
- <u>Graphic</u> stood out to be a special word as more successful project had that – but upon further interpretation, it was associated with graphic novel subcategory, which had an extremely high success rate

Hypothesis: There are specific keywords that make a project listing more appealing / less appealing than others





Wordcloud

We also did **Hierachical Clustering**, K-Means, K-Medoids (removed from code as there were no insights) but ultimately determined that there were no meaningful relationship between specific words and success of the project. In fact, many words were project-specific and clusters did not form according to categories / subcategories.

Problem Analysis

Model Development

Model Evaluation