



# **It's All Funds & Games Milestone Presentation**

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## Motivation (Refresher)

- We want to be able to predict the ultimate fate of a Kickstarter campaign
  - `final_status` column → failure = 0, success = 1
- Data comes from Kaggle (108,129 projects)
- Train-test Split: 66.6% - 33.3% (via `train_test_split` function from `sklearn.model_selection`)
  - Training Data: 72,446 projects
  - Testing Data: 35,683 projects



# Feature Engineering / Data Integration

- Here are some functions that we found very useful:
  - `to_datetime` (pandas library): converts unix time (1/1/1970) to datetime object
    - `created_at`, `launched_at`, `state_changed_at`, `deadline`, ...
  - `get_dummies` (pandas library): dummy codes categorical data
    - `country`, `launch_hour`, `launch_day`, `launch_month`, `deadline_day`, ...
- Expected Duration = `deadline - launched_at`
- Actual Duration = `state_changed_at - launched_at`



# Data Transformation Chart (Original)

Columns:

```
['name' 'desc' 'goal' 'keywords' 'disable_communication' 'country'
 'currency' 'deadline' 'state_changed_at' 'created_at' 'launched_at'
 'backers_count' 'final_status']
```

Data Types:

name	object
desc	object
goal	float64
keywords	object
disable_communication	bool
country	object
currency	object
deadline	int64
state_changed_at	int64
created_at	int64
launched_at	int64
backers_count	int64
final_status	int64
dtype:	object
Shape:	
(108129, 13)	



# Data Transformation Chart (Intermediate)

Columns:

```
['name' 'desc' 'goal' 'keywords' 'disable_communication' 'country'  
 'backers_count' 'final_status' 'expected_duration' 'actual_duration'  
 'launch_year' 'launch_month' 'launch_day' 'launch_hour' 'deadline_year'  
 'deadline_month' 'deadline_day']
```

Data Types:

name	object
desc	object
goal	float64
keywords	object
disable_communication	bool
country	object
backers_count	int64
final_status	int64
expected_duration	int64
actual_duration	int64
launch_year	int64
launch_month	int64
launch_day	int64
launch_hour	int64
deadline_year	int64
deadline_month	int64
deadline_day	int64

dtype: object

Shape:

(108129, 17)



# Data Transformation Chart ('Final')

Columns:

```
['name' 'desc' 'goal' 'keywords' 'disable_communication' 'backers_count'
'final_status' 'expected_duration' 'actual_duration' 'country_AU'
'country_CA' 'country_DE' 'country_DK' 'country_GB' 'country_IE'
'country_NL' 'country_NO' 'country_NZ' 'country_SE' 'country_US'
'launch_hour_0' 'launch_hour_1' 'launch_hour_2' 'launch_hour_3'
'launch_hour_4' 'launch_hour_5' 'launch_hour_6' 'launch_hour_7'
'launch_hour_8' 'launch_hour_9' 'launch_hour_10' 'launch_hour_11'
'launch_hour_12' 'launch_hour_13' 'launch_hour_14' 'launch_hour_15'
'launch_hour_16' 'launch_hour_17' 'launch_hour_18' 'launch_hour_19'
'launch_hour_20' 'launch_hour_21' 'launch_hour_22' 'launch_hour_23'
'launch_day_1' 'launch_day_2' 'launch_day_3' 'launch_day_4' 'launch_day_5'
'launch_day_6' 'launch_day_7' 'launch_day_8' 'launch_day_9'
'launch_day_10' 'launch_day_11' 'launch_day_12' 'launch_day_13'
'launch_day_14' 'launch_day_15' 'launch_day_16' 'launch_day_17'
'launch_day_18' 'launch_day_19' 'launch_day_20' 'launch_day_21'
'launch_day_22' 'launch_day_23' 'launch_day_24' 'launch_day_25'
'launch_day_26' 'launch_day_27' 'launch_day_28' 'launch_day_29'
'launch_day_30' 'launch_day_31' 'launch_month_1' 'launch_month_2']
```



# Classifier Evaluation

Model	Accuracy	Precision	Recall	F-1 Score	AUC
DT (Gini)	0.75985	0.49497	0.67057	0.56954	0.72907
LogReg	0.80063	0.44643	0.86852	0.58973	0.82787
RF (Gini)	0.85387	0.79193	0.76212	0.77674	0.83095
RF (Entropy)	0.85135	0.78905	0.75781	0.77312	0.82806
NB (Bernoulli)	0.62870	0.46651	0.42805	0.44645	0.58235
NB (Gaussian)	0.49995	0.87322	0.37894	0.52852	0.61133



# Progress

- Right now, we have classifiers built for the data **without** the columns that include text
- Ultimately, we want to cluster the projects into categories based on the text columns (name, description, keywords)
- We can then use the cluster/category as another feature
  - We expect to have to use stop-words, stemming, tokenization, and tf-idf
  - Tf-idf = “term frequency - inverse document frequency” : measures how significant a word is document or collection of documents