



Early Stage Project Success Measurement

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01

Design



Design

- **Why?**

1. Rapid changes in technology
2. Entrepreneurship is the mainstream

- **Goal:**

Determine if an idea/project is worth pursuing or not (success or fail).

Design

- **Who?**

Two main category of beneficiaries would use this project:

1. **Entrepreneurs:** To assess the quality of their idea
2. **Investors:** To determine what startups to invest in

- **How?**

Measure how likely are the users to pay for it. Success if the project achieve at least the financial goal or failure otherwise

02

Data & Algorithms



Data & Algorithms

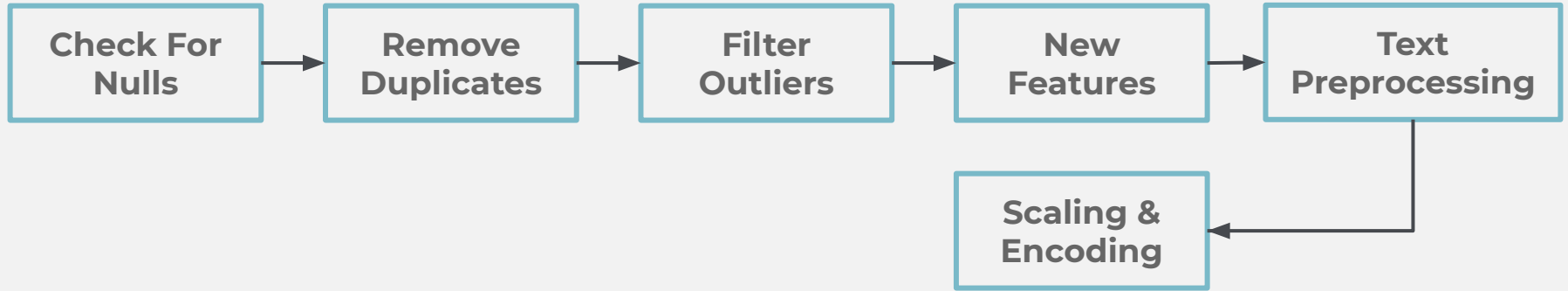
- **Data**

1. Kickstarter dataset ([Kaggle](#))
2. Contains 13 columns
3. 378,661 projects → After cleaning (124,235)

Data & Algorithms

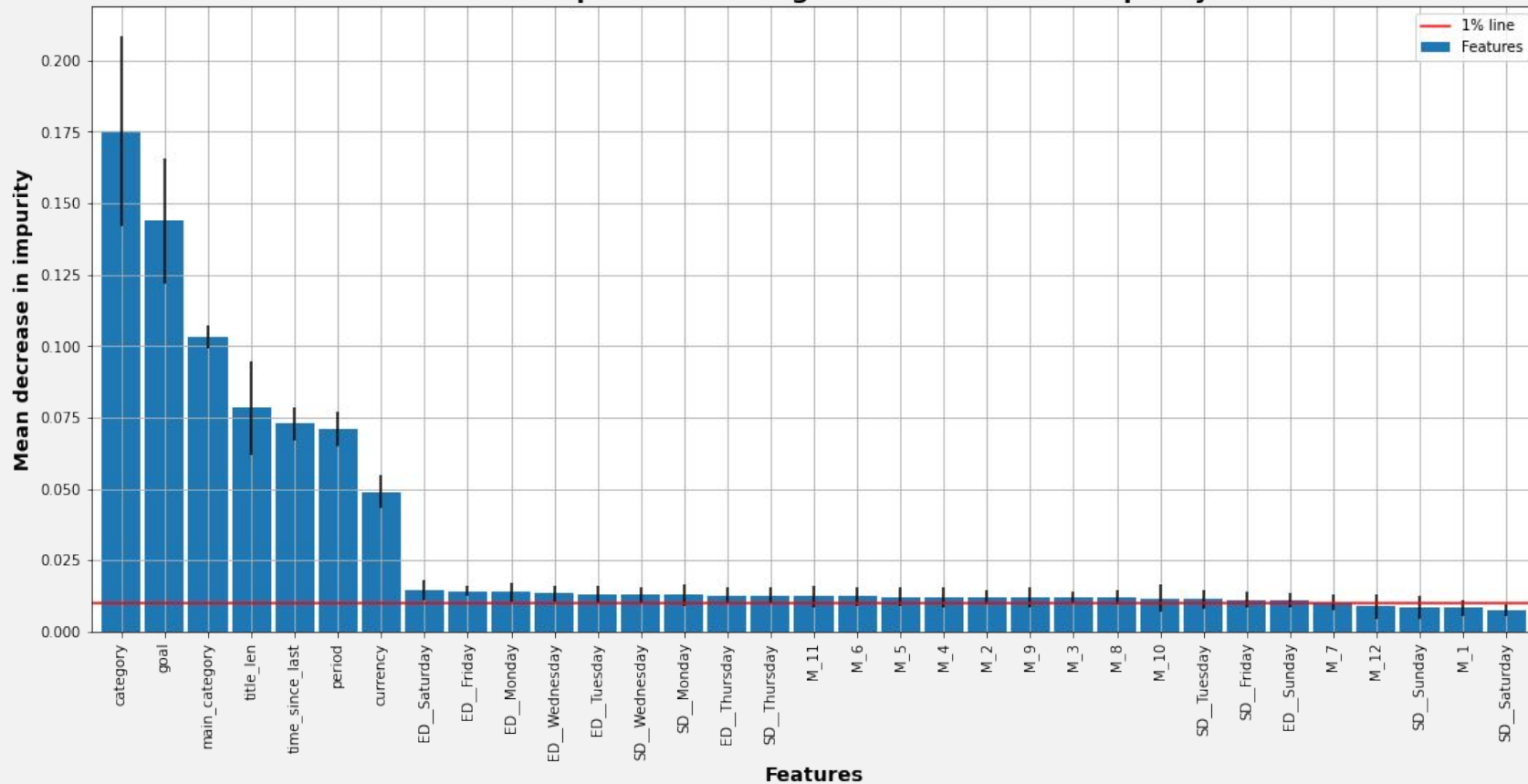
- **Preprocessing /Features Engineering:**

- The Preprocessing pipeline



Data & Algorithms

Feature Importances using Mean Decrease in Impurity



Data & Algorithms

- **Algorithms:**

Since it is a classification problem, several models were tested:

1. Classical Models (Logistic Regression & Support Vector Machine)
2. Ensemble **Bagging** Models (Random Forest)
3. Ensemble **Boosting** Models (Gradient Boosting)
4. Ensemble **Stacking** Models (Bert + Gradient Boosting)
5. Deep Learning Sequence Models (Bidirectional LSTM)
6. Pre-trained Models (Bert)

Data & Algorithms

Metrics	Logistic Regression	Support Vector Machine	Random Forest	Gradient Boosting	Bert + Gradient Boosting	Bi- LSTM	Bi- LSTM + NN	Bert
Accuracy	0.7445	0.7442	0.7447	0.7512	0.7505	0.7334	0.2666	0.7304
Precision	0.5664	0.5946	0.5539	0.5939	0.5905	0	0.2666	0.4838
Recall	0.1928	0.1375	0.2348	0.2210	0.2199	0	1	0.1647
F1	0.2877	0.2234	0.3298	0.3221	0.3204	-	-	-
AUC	0.7185	0.7184	0.7190	0.7384	0.7381	0.5	0.5	0.6218



Tools



Tools

1. **Data Processing:**

Pandas, and Numpy

2. **Modelling:**

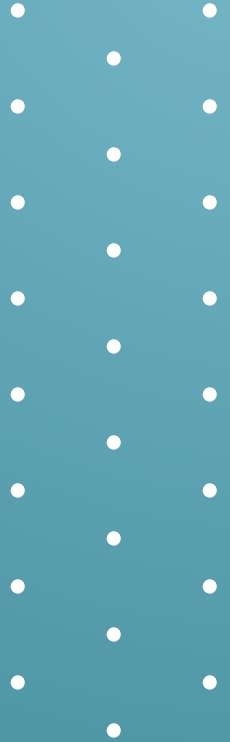
SciKit-Learn, PyTorch, TensorFlow/Keras, and Pre-trained models (Bert & GloVe)

3. **Visualization:**

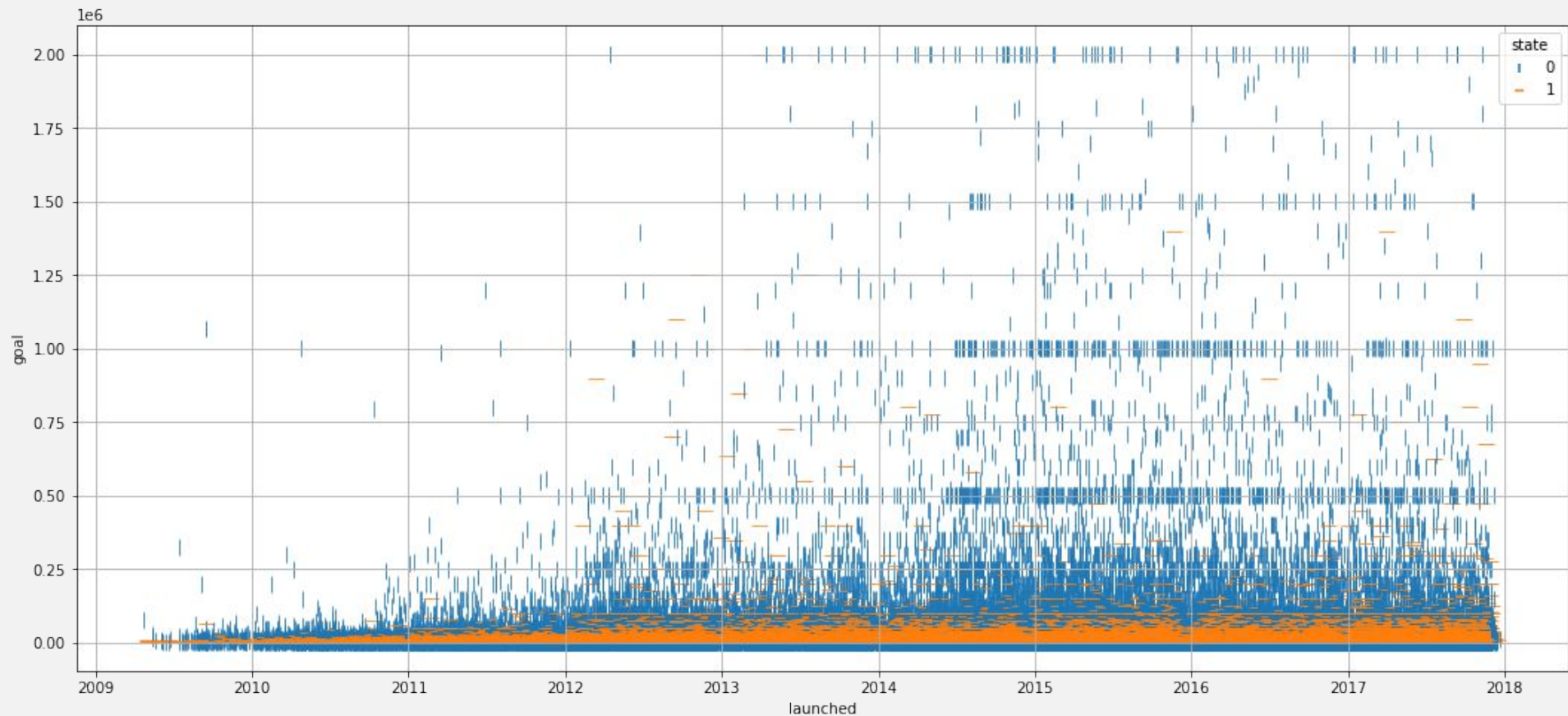
Matplotlib, Seaborn, and Google Colab

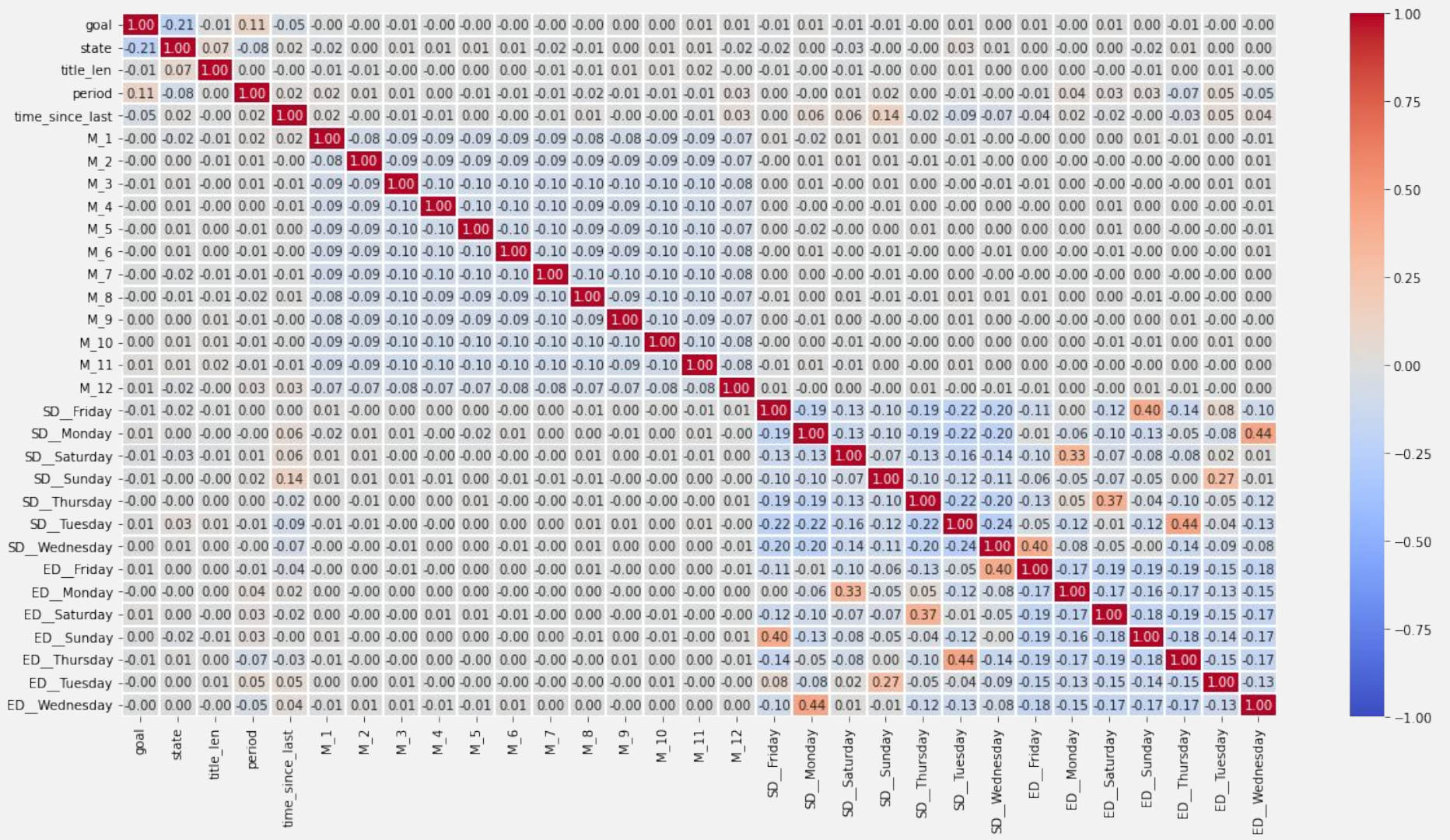
04

Insights & Conclusion



Insights & Conclusion





Insights & Conclusion

Insights:

1. **Model Range of Prediction:** $(5,000 \leq \text{Goal} \leq 2,000,000)$
2. **Best Dates:**
 - (Launch day: Tuesday)
 - (Launch month: October)
 - (Deadline day: Thursday)
3. **Best Categories:**
 - Music
 - Theater
4. **Worst Categories:**
 - Technology
 - Food
 - Film & Video

Insights & Conclusion

Prospective:

1. **Data is not sufficient:**


- Bias models → more complex which needs more features
- Project description/Images
- Unifying the currency of goal

2. **Web presence:**

- Integrated API / Stand alone website

3. **Utilizing more GPUs & RAMs:**

- Investigate more transformers/Pre-trained models



Thank You,
Any Questions?