

LendingClub

Building New Services through Predictive Modeling



I will walk you through three core talking points.

Problem Statement



Determine a good and bad loan before and during the
LendingClub loan approval process

First stop – what is the problem that we are trying to solve today.

Business Value



1. Better predictive capabilities
2. Better investors' outcome
3. More innovative products

Second stop – let's talk about business value of the problem we are trying to solve.

Better Predictive and Investor Outcome

We are able to create a better predictive outcome of loan default.

As a result, we will be able to attract more investors interest because they have higher confidence to invest.

More Innovative Products

Through understanding LendingClub's domain problem, and with predictive modeling modeling, we can design forward new services for LendingClub.

Methodology



- OSEMN Data Science Workflow
 - Basic Predictive Model
 - Deep Neural Network
- Design Sprint and Agile Development

I have breakdown my methodology into two parts. First, is the data science workflow where I leveraged on OSEMN as a guide post; and based on the workflow, I have started to obtain the data, scrub the data, explore the data, model the data, and interpret the data. In predictive modeling, I have used the basic predictive model, and deep neural network. Second, I have used the design sprint and agile development to attempt to build new services.

Results

The results based on basic predictive deep neural network are almost identical – swinging between 78% and 79% test accuracy

Basic Predictive Modeling

- | | |
|------------------------|---------|
| 1. Logistic Regression | : 78.5% |
| 2. Random Forest | : 78.5% |
| 3. XGBoost | : 77.2% |
| 4. SVM | : 78.5% |

Deep Neural Network Modeling

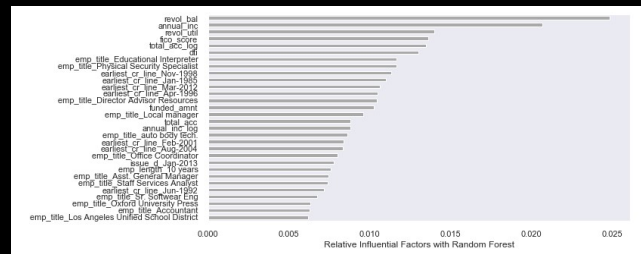
- | | |
|-------------|-----|
| 1. Layer 1: | 79% |
| 2. Layer 2: | 78% |
| 3. Layer 3: | 79% |



There is not much different between the basic, and more complicated deep neural network. If the basic predictive modeling have almost similar predictive power, then we should use the basic model to save effort and cost.

Results

Based on Random Forest – relative influential factors to predict bad and good loans.

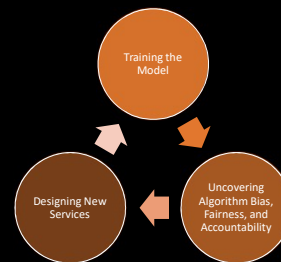


This is the relative importance / predictors for LendingClub to consider when building new algorithm, and new services

Conclusion

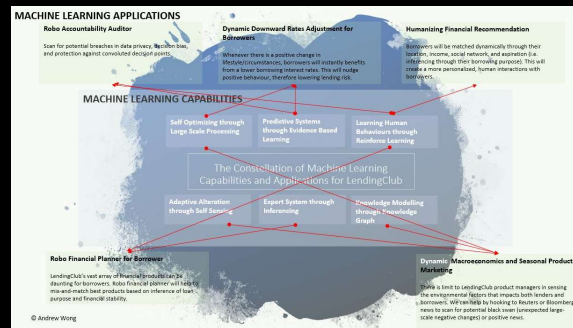
Product-Driven Machine Learning

The learning-feedback cycle
Constant feeding of updated datasets, as well as new type of datasets.



Conclusion

What Machine Learning capabilities that LendingClub can build based on the predictive modeling that I have built (no doubt it is not as simple as that!).



Based on the predictive modeling results, I have attempted to design forward new capabilities, and applications for LendingClub.

Future Work

1. Faster and more scalable processing in the cloud
2. Focus on predicting investors' risk behaviour



For future work. First, I'd like to run the dataset in the cloud to save time and effort. Second, the current problem statement focus on borrowers' behaviour. I'd like to shift my focus on investors' behaviour, especially on their risk management.



Q &A