

CS583A: Course Project

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1 Summary

[Problem descriptions:] I have participated in a inactive competition with late submission, the project is to identify if or not a customer will make a transaction or no. [Implementation:] Here I have compared and contrasted various methods:

- Logistic Regression
- Naive Bayes
- Simple Feed Forward neural network
- CNN

[Evaluation metric:] Submissions are evaluated on area under the ROC curve between the predicted probability and the observed target. [Score and ranking:] Have attached screenshots of the code, there was no ranking as this was a late submission

2 Problem Description

Problem. The problem is to classify if a customer will make a transaction or not based on 200 features and a binary target, it is a binary classification problem. This is an important factor for banks to help them promote and make understand why customers do not make transactions. The competition is at <https://www.kaggle.com/c/santander-customer-transaction-prediction/overview/evaluation>.

Data. The data are 200k x 202 training samples and the test is 200k x 201. The number of classes is 2. The training set is not well-balanced: $n_1 = 10.049\text{percent}$ and $n_0 = 89.951\text{percent}$.

Challenges. There are quite a few challenges in this dataset, first of all we need to identify the most important parameters and the correlations between these, there are quite a few things that I understood that I will talk about later, the dataset is imbalanced.

3 Solution

Model. The model we finally choose is a CNN with quite a few modifications

Layer (type)	Output Shape	Param #
conv1d_1 (Conv1D)	(None, 200, 600)	1800
flatten_1 (Flatten)	(None, 120000)	0
dense_1 (Dense)	(None, 1)	120001
Total params: 121,801		
Trainable params: 121,801		
Non-trainable params: 0		

Figure 1: CNN model

Implementation. We implement the CNN

We run the model using keras with tensorflow as backend, since I had tested 3 other models, I had to run this one on the cloud using the colab gpu by google. The other models were run locally on Razer blade 15 with rtx 2060 and i7 and 16 gb memory. The cnn took 1.5 hours to train.

Settings. The loss function is categorical binary-cross entropy. The optimizer is Adam. Here we use the CLR(Cyclic learning rate provided by the keras contrib package, it has a good range of lr's) Kernel-size =2, strides=2, sigmoid activation

The kernel and the strides are two to help us in picking the new encoded feature and the original together. We also use 600 filters Batch size=512

epochs =5

There can be a very good score given the resources to train, this was trained on very constrained conditions of memory and gpu.

Advanced tricks. Here based on a few discussion threads it was found that the test data had fake rows in it based on the frequency and similarity of the datapoints, so removed these, then based on another discussion thread, it was seen that all the parameters were independent of each other and if we added extra parameters by frequency encoding each of the 200 variables, this gave a large boost in the score.

Also to ensure the CNN which is very good at finding correlations between variables, we try to augment the data and feed it to the network in such a way that it always sees a different subset of the data.

Cross-validation. The parameters were tuned using a 5 fold cross-validation

4 Compared Methods

Logistic Regression. I have used the oldest model in the book for classification, use 10 fold CV with with lbfgs with 1500 iterations, the best AUC score at the end of the cross validation was 0.85

Layer (type)	Output Shape	Param #
input_12 (InputLayer)	(None, 200, 3)	0
dense_52 (Dense)	(None, 200, 32)	128
batch_normalization_25 (Batch Normalization)	(None, 200, 32)	128
dense_53 (Dense)	(None, 200, 8)	264
batch_normalization_26 (Batch Normalization)	(None, 200, 8)	32
flatten_12 (Flatten)	(None, 1600)	0
dense_54 (Dense)	(None, 1)	1601
Total params: 2,153		
Trainable params: 2,073		
Non-trainable params: 80		

Figure 2:

Naive Bayes. This is also a classic model in classification tasks, by using the GaussianNb from sklearn and fitting it to the data the AUC score was 0.89

Feed Forward neural network We use the keras for building and training this neural network. Below is the architecture of the data. Even here we remove the fake data and add extra features. The auc score from this was 0.90

Advanced tricks. We finally adopted the model as shown in figure 3 We applied the following tricks.

- Data augmentation. This helps maintain the independence of features and forces the CNN to look at each feature independently
- Adding extra features based on discussion threads, we encode the frequency to the data this helps create additional 200 columns.

The final score was 0.91

5 Outcome


Participated in a inactive with late submission but going by the scores it can be seen that the ranking would be in the top 20 percent. Again given the constraints of the memory and the gpu we could do a lot better on the CNN in Figure 3. There html files uploaded have some manual stops given the memory issues and the gpu issues I had to restart them, the score submitted is the best of them.

← → ↺ https://www.kaggle.com/c/santander-customer-transaction-prediction/submissions

Apps Hackers Can Remot... Kali Linux Howto's... Lecture 1 | Machine... (15) How can I lear... 2015 Victoria's Secr... Inbox (41,907) (11) What are some... (11) Adrián Lamo's... (11) How do I bypa...

Santander Customer Transaction Prediction

Can you identify who will make a transaction?

 Banco Santander · 8,802 teams · a month ago

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Your most recent submission

Name	Submitted	Wait time	Execution time	Score
submission.csv	2 minutes ago	1 seconds	2 seconds	0.91773

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Your final score may not be based on the same exact subset of data as the public leaderboard, but rather a different private data subset of your full submission — your public score is only a rough indication of what your final score is.

You should thus choose submissions that will most likely be best overall, and not necessarily on the public subset.

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kaggle competitions submit -c santander-customer-transaction-prediction -f submission.csv -m "Message"

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2 submissions for [VardaanKhanted](#)

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Submission and Description	Private Score	Public Score	Use for Final Score
submission.csv 2 minutes ago by VardaanKhanted add submission details	0.91772	0.91773	<input type="checkbox"/>

Figure 3: Caption