# 7th place solution - HC + Ridge

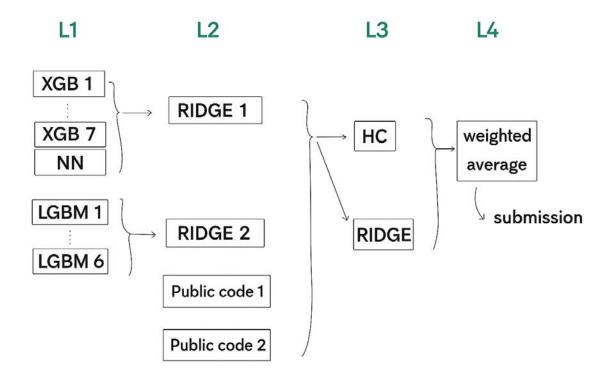
Thank you for the past month!

It was an honor to achieve this ranking in a competition I participated in wholeheartedly from start to finish.

In every competition, I make it a point to explore and deeply experiment with new techniques. In this competition, I challenged myself with multi-level ensembling, something I had never tried before — up until now, I had only done simple weighted averaging of multiple models at most.

Below, I will share my 1. model, 2. reflections, 3. acknowledgments.

### 1. model



# L1

For the first 20 days or so, I focused on implementing base models for ensembling. To increase diversity, I varied the following 3 aspects, tuned the remaining hyperparameters for optimal performance, and kept only the models that contributed to improving the overall prediction.

1. how many original datas added to the training data

- 2. treat numerical features as cat or int
- 3. reg alpha, reg lambda params

As a result, 7 XGB models( $\rightarrow$ RIDGE1) and 6 LGBM models( $\rightarrow$ RIDGE2) remained. I also made NN model (refer to <u>here</u>, changed some points like num of folds).

#### L2

I used 2 public codes( one two ) in order to improve the score.

I pruned the ensemble candidates that used approaches different from mine as much as possible.

#### L3

At this layer, I ultimately used both HC and RIDGE.

I also experimented with stacking models like NN and XGB, but they didn't help improve the score at all, so I ended up abandoning them.

Although I was initially hesitant to use HC at this layer, the cross-validation score for the final submission was clearly better, so I decided to go with it.

#### L4

HC(L3)

OOF: 0.38396, private LB: 0.38460

RIDGE(L3)

CV: 0.38368, private LB: 0.38449

→last submission(L4)

CV: 0.38412, private LB: 0.38486

#### 2. reflection

## positive aspect

Reflecting on the previous competition and heeding the warnings from Discussions, I was able to fully trust my CV this time.

To be honest, since I'm still relatively new to Kaggle, I used to get overly excited or discouraged by the public LB, but in the final week, I managed to stay sane:)

## negative aspect

First of all, I completely forgot that ridge regression has a hyperparameter.

It happened to work well with the initial setting, so I left it as is and got caught up in building the stacking model.

I only realized this just now after looking at other people's solutions...

Also, I had been naming all my submission files the same (ensemble\_submission.csv), so I lost track of which submission was which.

As a result, I ended up taking a bit of a gamble for the final submission.

Miraculously, I submitted the one with the best CV score, but I definitely need to reflect on this.

# 3. acknowledgments

I was able to come this far thanks to all the things I learned from everyone.

While I can't mention every single person here, I'd like to express my gratitude to the following individuals.

@siukeitin, @tilii7, @richardjana, @masayakawamata @cdeotte, @robschieber, @paperxd, @mahoganybuttstrings @act18l, @gauravduttakiit, @ravi20076, @yunsuxiaozi, @gowthamdd thanks!!