ALLSTATE CLAIMS SEVERITY

CREATE AN ALGORITHM ACCURATELY PREDICTING CLAIMS SEVERITY

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ABOUT THIS PROJECT

- This is a previous recruitment prediction competition held by Kaggle (an online community of data scientists) and Allstate Insurance.
- Allstate invited machine learning practitioners to show off their creativity and flex their technical chops by creating an algorithm which accurately predicts claims severity.
- Submissions are evaluated on the mean absolute error (MAE) between the predicted loss and the
 actual loss.

DATA DESCRIPTION

- Each row in this dataset represents an insurance claim. Competitors must predict the value for the 'loss' column. Variables prefaced with 'cat' are categorical, while those prefaced with 'cont' are continuous.
- File descriptions
 - o train.csv the training set
 - test.csv the test set. You must predict the loss value for the ids in this file.
 - o sample_submission.csv a sample submission file in the correct format

■ Train Dataset: 188,318 rows, 132 columns

	id	cat1	cat2	cat3	cat4	 cont11	cont12	cont13	cont14	loss
0	1	A	В	A	В	 0.569745	0.594646	0.822493	0.714843	2213.18
1	2	Α	В	Α	Α	 0.338312	0.366307	0.611431	0.304496	1283.60
2	5	A	В	A	A	 0.381398	0.373424	0.195709	0.774425	3005.09
3	10	В	В	Α	В	 0.327915	0.321570	0.605077	0.602642	939.85
4	11	A	В	A	В	 0.204687	0.202213	0.246011	0.432606	2763.85

Test Dataset: 125,546 rows, 131 columns

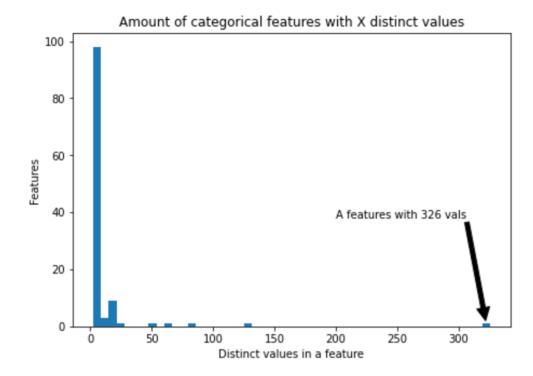
	id	cat1	cat2	cat3	cat4	• • •	cont10	cont11	cont12	cont13	cont14
0	4	A	В	A	A		0.38016	0.377724	0.369858	0.704052	0.392562
1	6	Α	В	A	В		0.60401	0.689039	0.675759	0.453468	0.208045
2	9	A	В	A	В		0.30529	0.245410	0.241676	0.258586	0.297232
3	12	A	Α	A	A		0.31480	0.348867	0.341872	0.592264	0.555955
4	15	В	A	A	A		0.50556	0.359572	0.352251	0.301535	0.825823

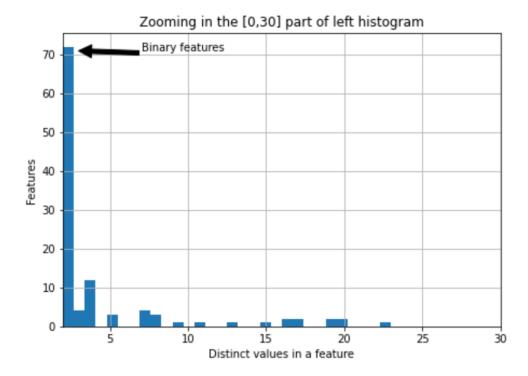
Train Dataset:

First 20 columns: ['id', 'cat1', 'cat2', 'cat3', 'cat4', 'cat5', 'cat6', 'cat7', 'cat8', 'cat9', 'cat10', 'cat11', 'cat12', 'cat13', 'cat14', 'cat15', 'cat16', 'cat17', 'cat18', 'cat9', 'cat10', 'cat11', 'cat12', 'cat13', 'cat14', 'cat15', 'cat17', 'cat18', 'cat9', 'cat10', 'cat11', 'cat11' Last 20 columns: ['cat112', 'cat113', 'cat114', 'cat115', 'cat116', 'cont1', 'cont2', 'cont3', 'cont4', 'cont5', 'cont6', 'cont7', 'cont8', 'cont9', 'cont10', 'cont11', 'cont12', 'cont9', 'cont10', 'cont10' id cont1 cont2 cont3 cont4 cont6 cont7 cont8 cont9 cont10 cont11 cont12 188318.000000 188318.000000 188318.000000 count 188318.000000 188318.000000 188318.000000 188318.000000 188318.000000 188318.000000 188318.000000 188318.000000 188318.000000 188318.000000 294135.982561 0.493861 0.507188 0.498918 0.491812 0.487428 0.490945 0.484970 0.486437 0.485506 0.498066 0.493511 0.493150 mean 169336.084867 0.187640 0.207202 0.202105 0.211292 0.209027 0.205273 0.178450 0.199370 0.181660 0.185877 0.209737 0.209427 std 1.000000 0.000016 0.001149 0.002634 0.176921 0.281143 0.012683 0.069503 0.236880 0.000080 0.000000 0.035321 0.036232 min 147748.250000 0.358319 0.327354 0.281143 0.336105 0.358970 0.364580 0.346090 0.336963 0.350175 0.312800 0.310961 0.311661 294539.500000 0.475784 0.555782 0.527991 0.452887 0.422268 0.440945 0.438285 0.441060 0.441450 0.461190 0.457203 0.462286 440680.500000 0.623912 0.681761 0.634224 0.652072 0.643315 0.655021 0.591045 0.623580 0.566820 0.614590 0.678924 0.675759 0.984975 0.862654 0.944251 0.954297 0.983674 0.997162 1.000000 0.980200 0.995400 0.994980 0.998742 587633.000000 0.998484

- Datasets have been preprocessed since all the continues variables have been reduced to [0,1] and means of them are around 0.5. Therefore, this is a feature dataset.
- No missing values.

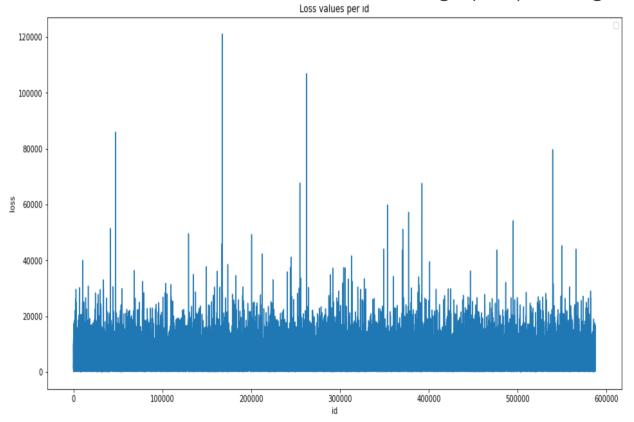
- Categorical: I 16 features
- Continuous: 14 features
- A column of int64: ['id']

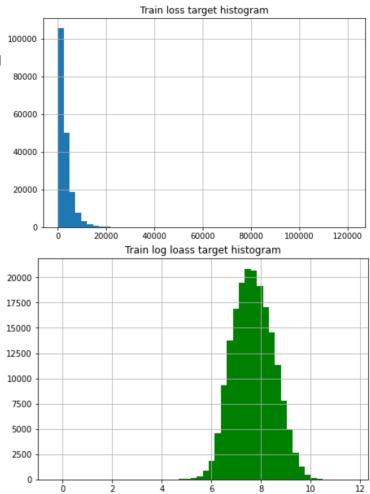




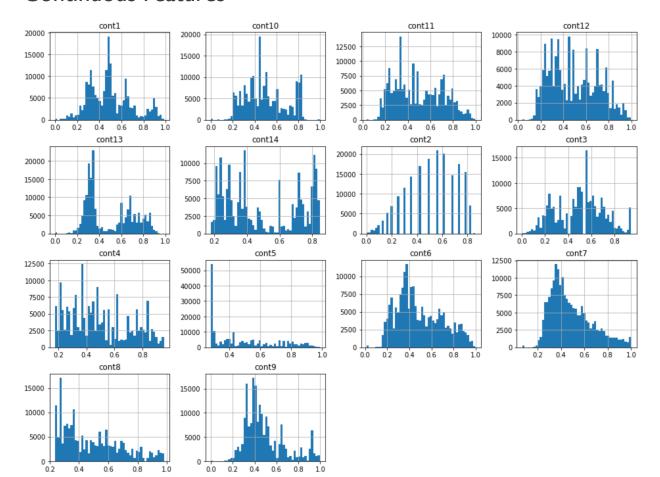
Target Feature: Loss Value

Since the skewness of the loss value is high (3.67), use log transform

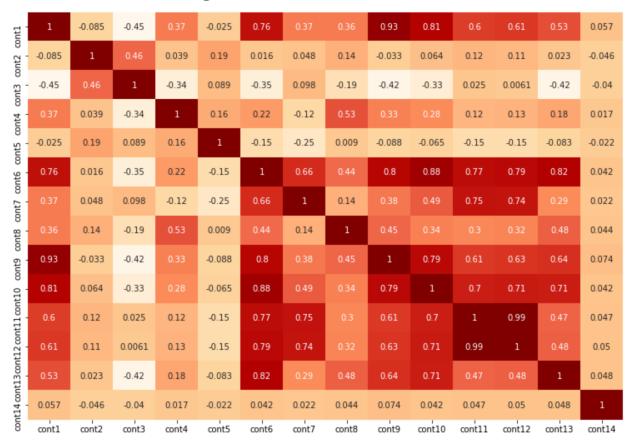




Continuous Features



Correlations among features



- 0.8

- 0.6

- 0.4

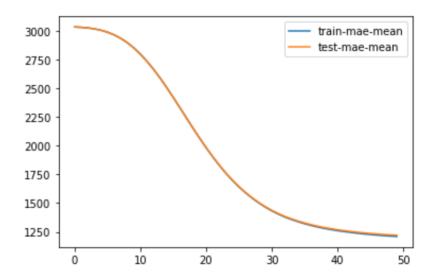
- 0.2

- 0.0

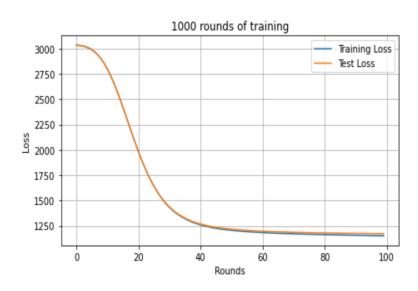
- -0.2

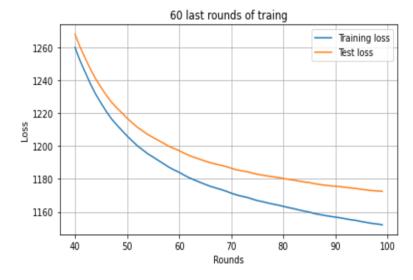
- -0.4

- XGBoost Regression
- Cross validation
 - 50 trees: cv score = 1220.14



100 trees: cv score = 1172.46 (overfitting risk)



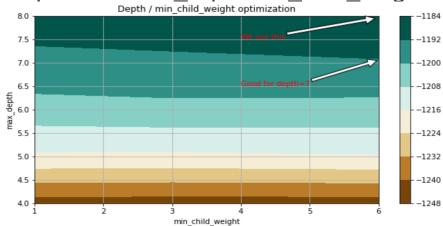


Grid Search

Step I : Build base model: eta=0.1, colsample_bytree=0.5, subsample=0.5, max_depth=5, min_child_weight=3, num_boost_round=50

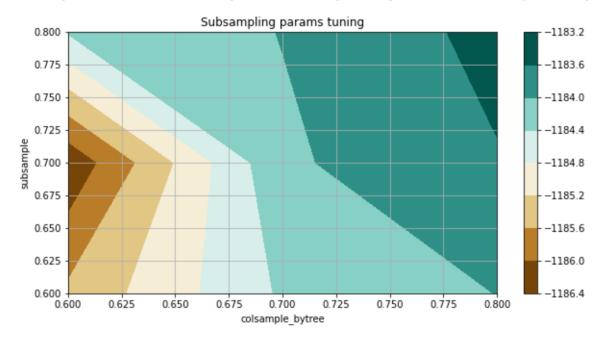
```
train-mae-mean
                   1208.575903
                      2.065637
train-mae-std
train-rmse-mean
                      0.558723
train-rmse-std
                      0.000888
test-mae-mean
                   1217.096240
                     11.171228
test-mae-std
test-rmse-mean
                      0.562311
                      0.002872
test-rmse-std
Name: 49, dtype: float64
```

Step 2: Tune max_depth, min_child_weight



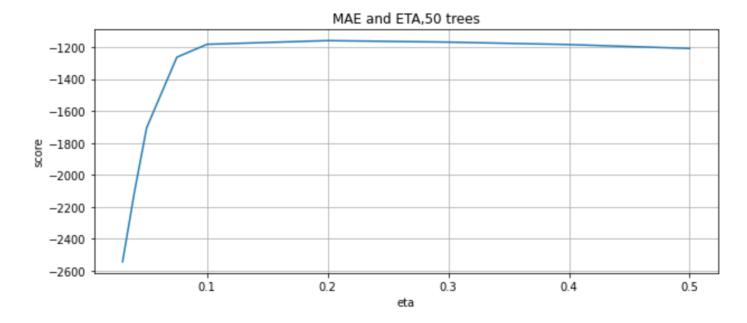
Grid Search

- Step 3: Tune gamma to lower overfitting risk
- Step 4: Tune subsample, colsample_bytree to change sampling

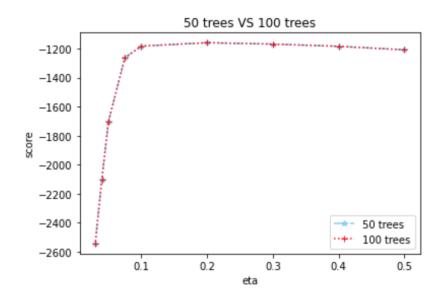


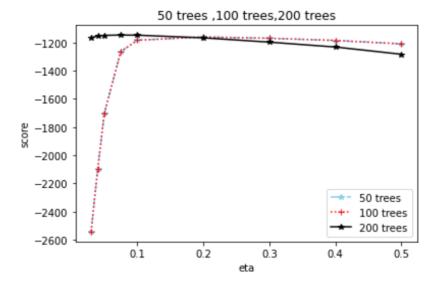
Grid Search

Step 5: Tune eta



Grid Search





- Find the best combination of trees and learning rate: eta = 0.075 num_boost_round = 200
- Make predictions
 - o array[7.3063416, 7.6374493, 9.159702, ..., 7.842589, 6.9370203, 7.945583]