

Approach Machinehack analytics olympiad 2022

Create a machine learning model to help an insurance company understand which claims are worth rejecting and the claims which should be accepted for reimbursement..

- Basic exploratory data analysis using pandas, matplotlib, seaborn packages.
- Data pre-processing
 - Column name lower case
 - Feature Engineering
 - Create a binary feature if a person having child without a marriage.
 - get a categorical columns percentage
 - numerical groupby numerical summary
- The final features for the model
 - 1_age
 - 2_gender
 - 3_driving_experience
 - 4_education
 - 5_income
 - 6_credit_score
 - 7_vehicle_ownership
 - 8_vehicle_year
 - 9_married
 - 10_children

- 11_postal_code
- 12_annual_mileage
- 13_speeding_violations
- 14_duis
- 15_past_accidents
- 16_outcome
- 17_type_of_vehicle
- 18_data
- 19_not_married_hv_child
- 20_id_age_percentage
- 21_id_gender_percentage
- 22_id_driving_experience_percentage
- 23_id_education_percentage
- 24_id_income_percentage
- 25_id_vehicle_ownership_percentage
- 26_id_vehicle_year_percentage
- 27_id_married_percentage
- 28_id_children_percentage
- 29_id_type_of_vehicle_percentage
- 30_id_not_married_hv_child_percentage
- 31_id_annual_mileage_percentage
- 32_id_speeding_violations_percentage
- 33_id_duis_percentage
- 34_id_past_accidents_percentage
- 35_age_credit_score_median

- 36_age_credit_score_mean
- 37_age_credit_score_max
- 38_age_credit_score_min
- 39_age_annual_mileage_median
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- 48_age_duis_mean
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- 50_age_duis_min
- 51_age_past_accidents_median
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- 57_gender_credit_score_max
- 58_gender_credit_score_min
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- 60_gender_annual_mileage_mean

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- 70_gender_duis_min
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- 78_driving_experience_credit_score_min
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- 81_driving_experience_annual_mileage_max
- 82_driving_experience_annual_mileage_min
- 83_driving_experience_speeding_violations_median
- 84_driving_experience_speeding_violations_mean
- 85_driving_experience_speeding_violations_max

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- 185_married_speeding_violations_max

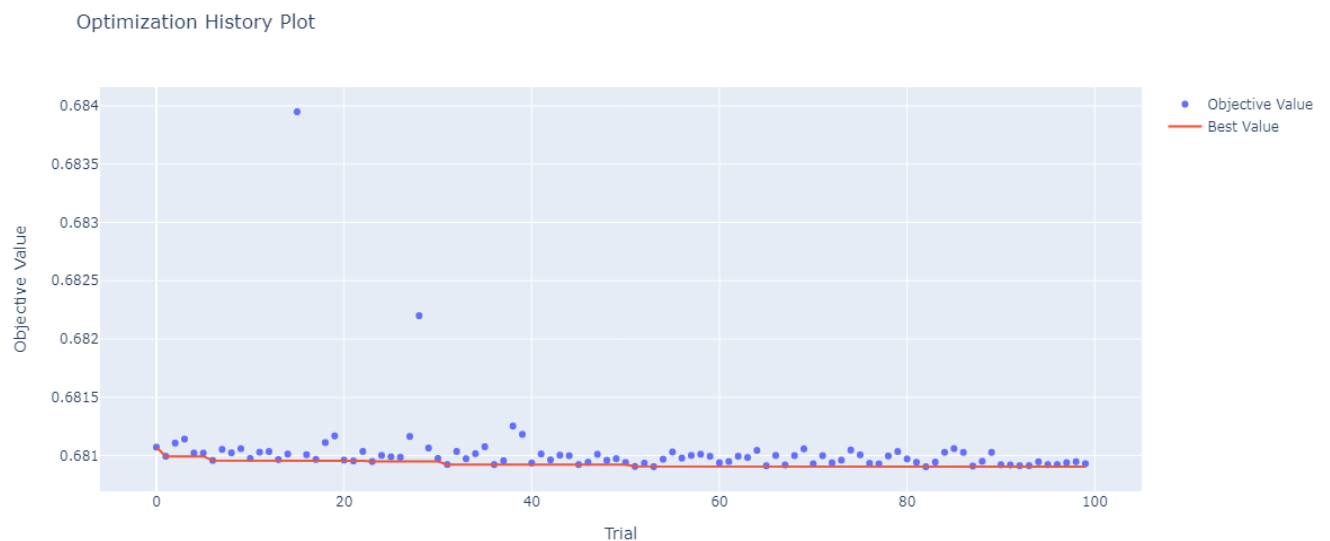
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- 211_not_married_hv_child_past_accidents_median
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- 229_children_duis_max
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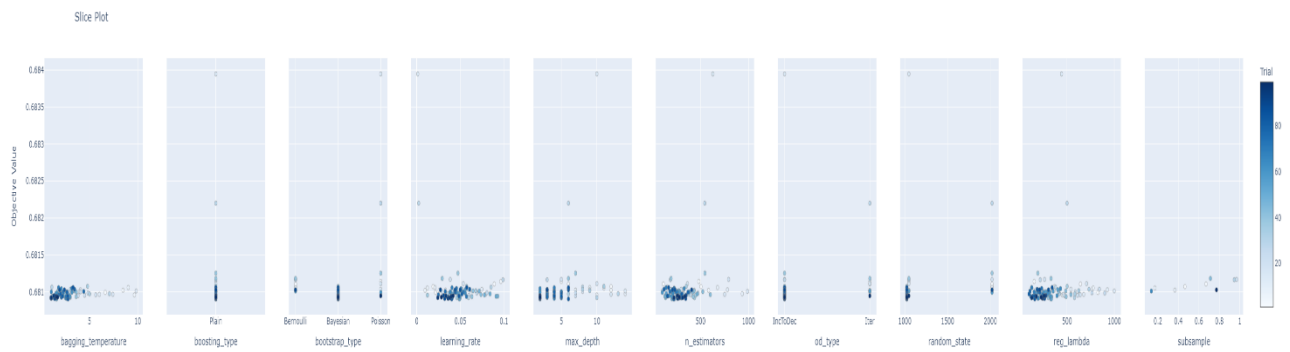
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- 254_type_of_vehicle_past_accidents_min

- Created catboost classifier model and tuned hyperparameters by using optuna framework. Model evaluated with Logloss. After 100 trials,
 - The best score is 0.68106
 - The best hyperparemeters are,

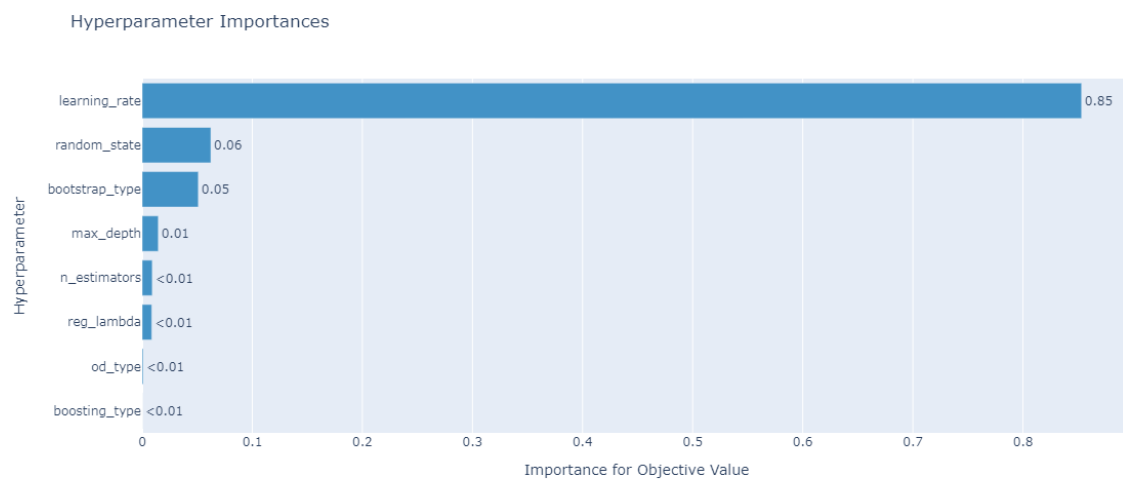

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'random_state': 1024,
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'bootstrap_type': 'Bayesian',
'bagging_temperature': 2.7779687356027973}
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- Visualizing the Optimization History - Explains the best score at each trials.



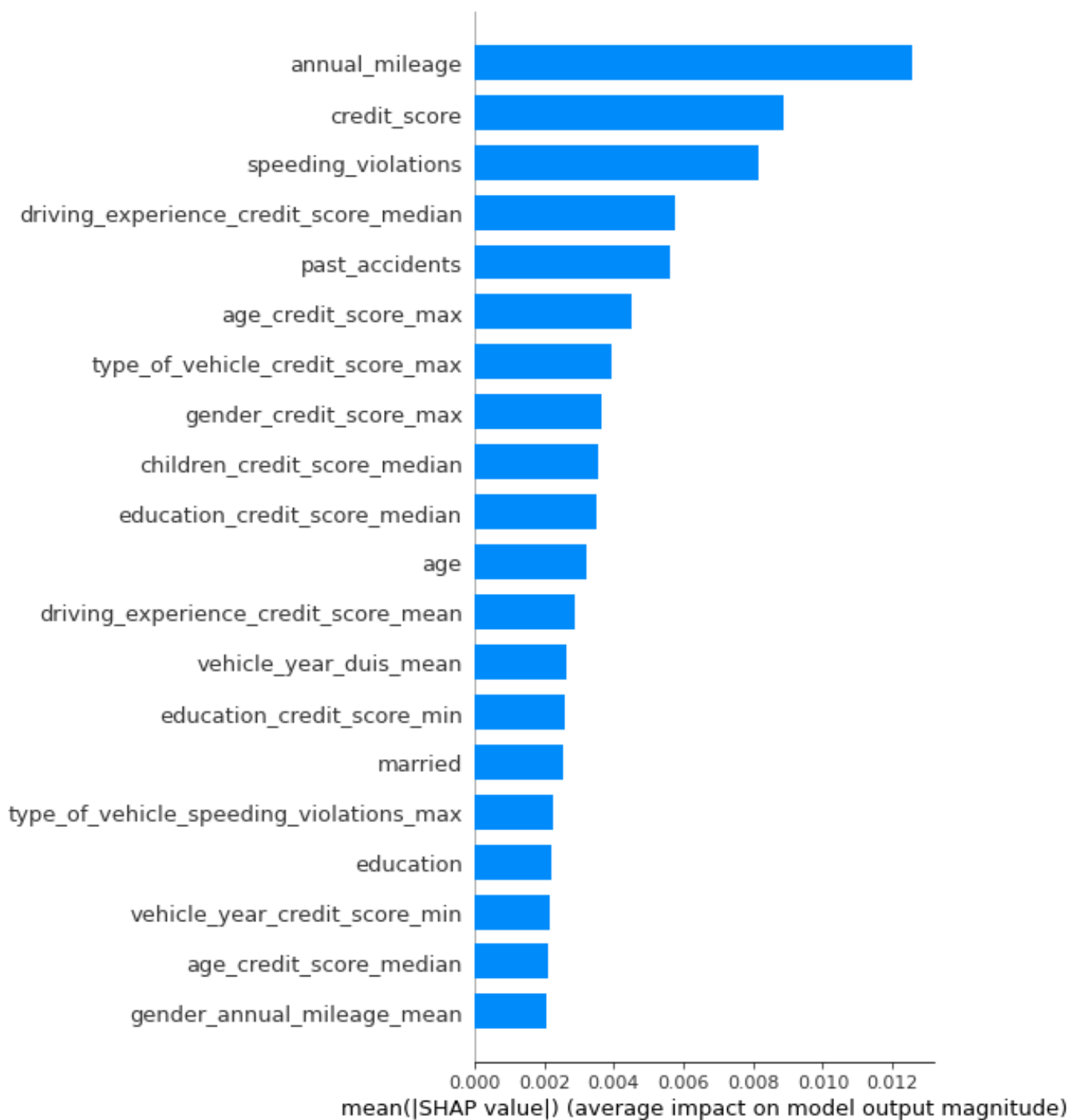
- Visualizing High-dimensional Parameter Relationships



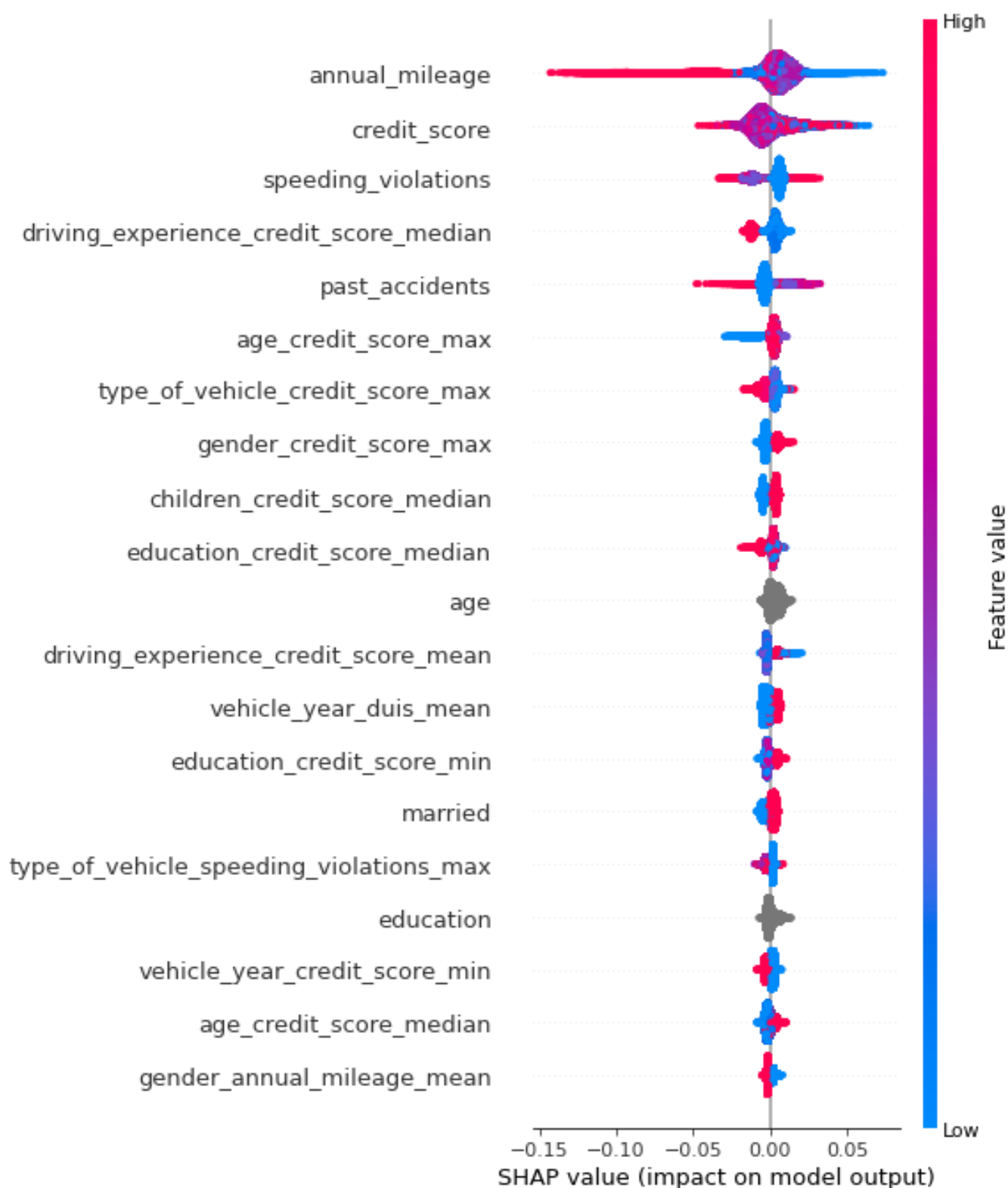
- Visualizing Parameter Importances



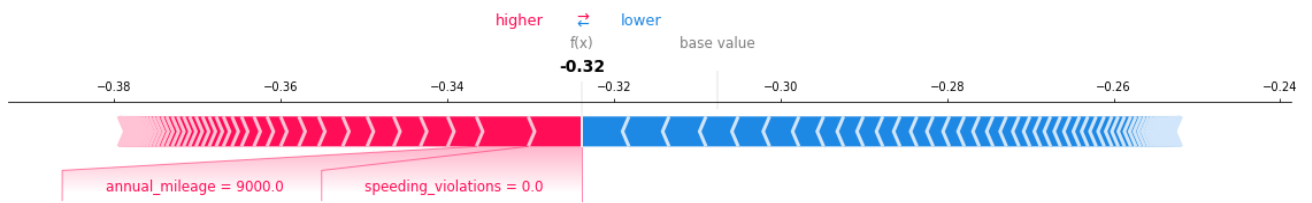
- Catboost SHAP feature importances plot



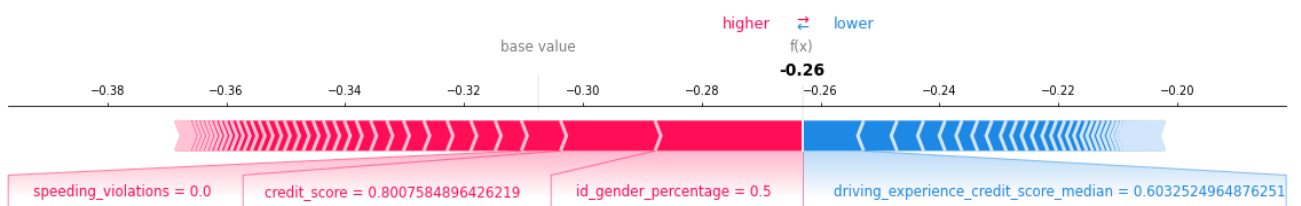
- Catboost SHAP top features impact the model



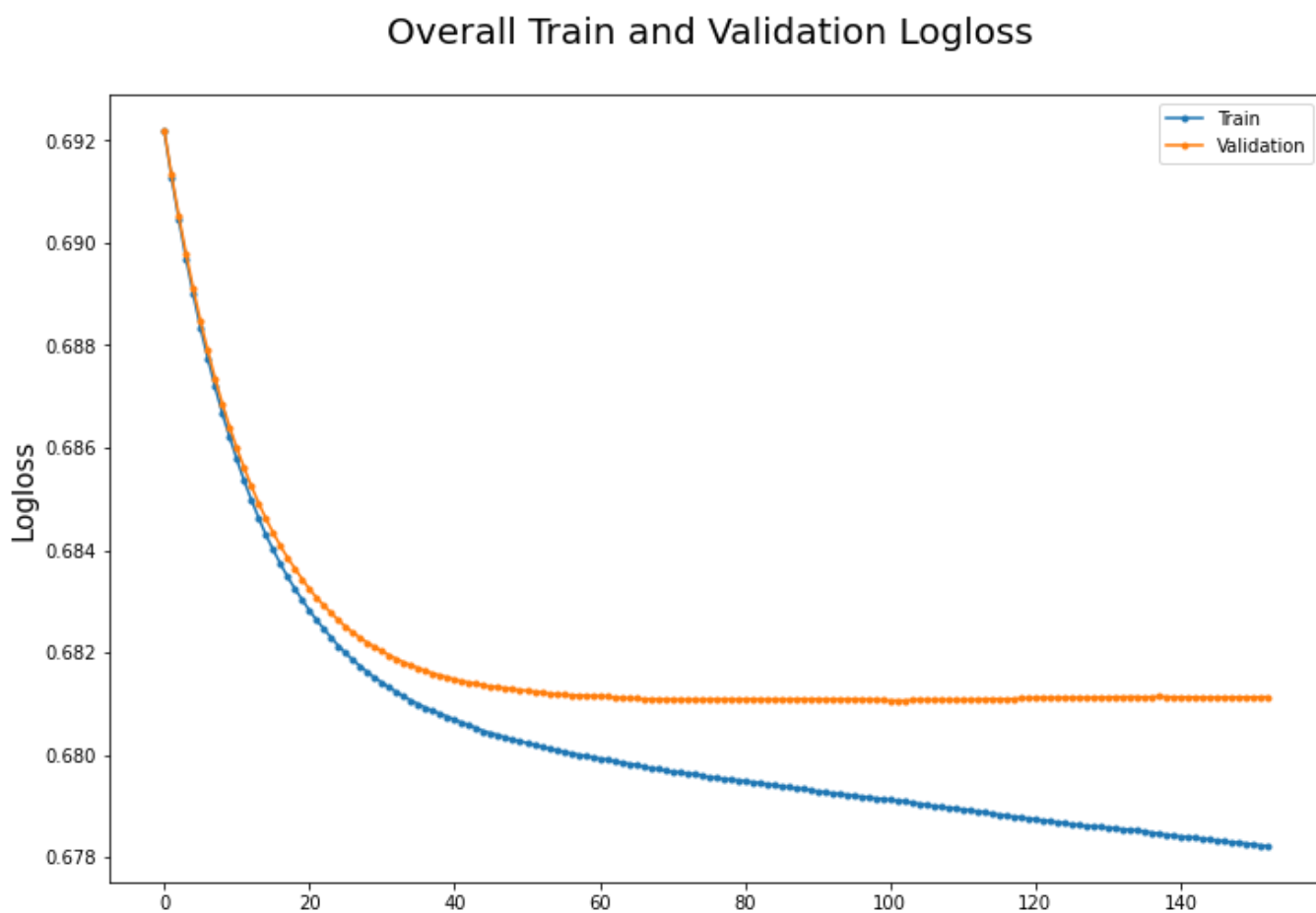
- Top feature influences for class 1



- Top feature influences for class 0



- Overall Train and Validation Logloss



- Train logloss: 0.67936 , Validation logloss: 0.68106
- Final competition score is 0.68081