Machine Learning for Shot Classification on the DIII-D Tokamak

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Introduction

- Plasma zoology
 - \circ Unclear distinctions
 - \circ ITER mode optimization
- Categorization
 - Human fallibility
 - Machine learning applications

Background: Plasma Modes

- Quiescent high confinement mode (QH mode)
 - Wide pedestal
 - Super-H
 - ITER baseline scenario
 - Unspecified
- Resonant magnetic perturbation (RMP mode)
 - ITER steady state
 - ITER baseline scenario
 - Unspecified

- Negative triangularity mode (Neg-D)
 - Inner wall limited
 - Diverted
- Low confinement mode (L mode)
 - Improved confinement
 - Enhanced D-Alpha
 - High power
 - Standard
 - Inner wall limited
- Edge localized mode (ELM)

Background: Machine Learning Models

- Tree classification
 - Decision Tree
 - Random Forest
 - Extreme Trees
 - Extra Trees
- Other classifiers
 - Gradient Boosting
 - Bagging
 - Gaussian Naive Bayes
 - Quadratic Discriminant Analysis
- Stacking ensemble



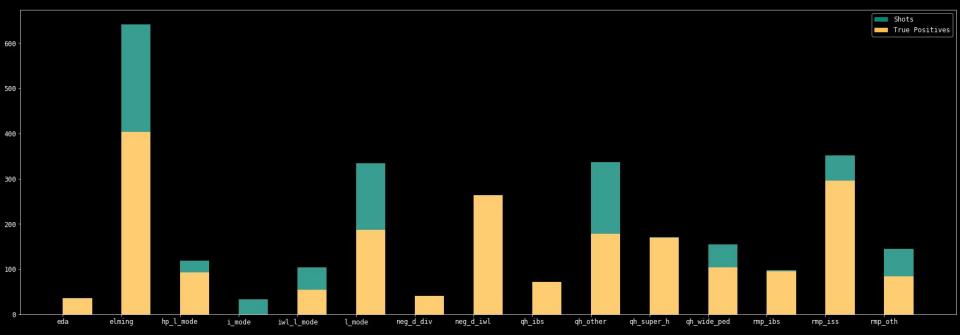
Methodology

• Data

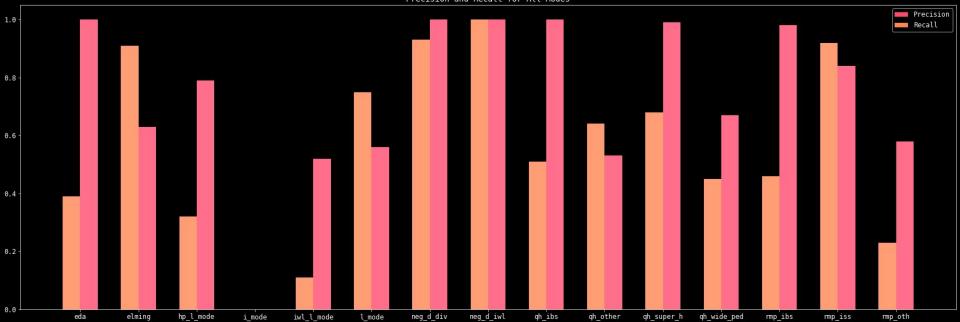
- Hand-classified database
- Single mode shots
- Signals: q0, q95, βN, toroid B field strength, upper & lower triangularity, I30 coil current, topology
- 190 data points
- Model selection
 - Tested 41 models on 80-20 train-test split
 - Eliminated low performance and failed convergence models
 - Tested different ensemble configurations
 - Selected highest performance ensemble
- Class imbalance problem
 - GAN
 - ADASYN
 - SMOTE

Results

True Positive Distribution in Testing Dataset

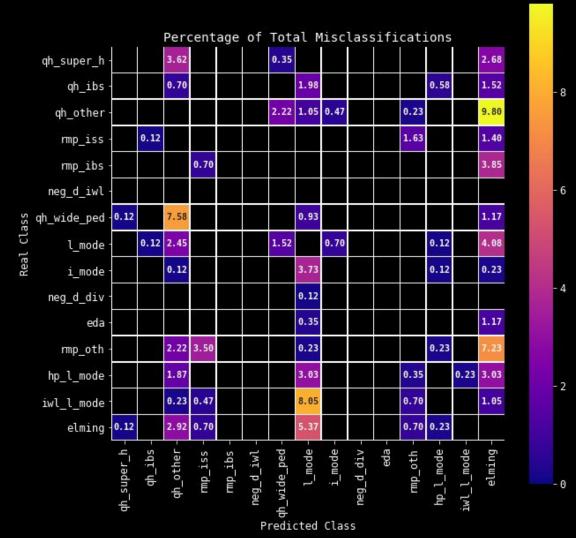


Results



Precision and Recall for All Modes







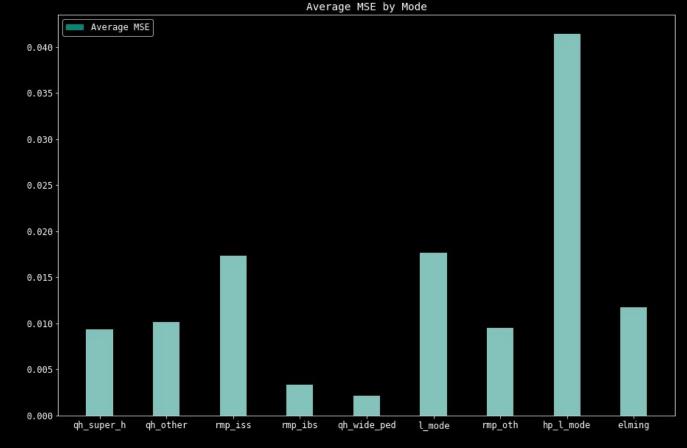
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Applications

 Profile prediction validation

QH Super H	3
QH (other)	26
RMP ISS	18
RMP IBS	2
QH Wide	3
Pedestal	
L Mode	62
RMP (other)	3
HP L mode	8
ELMing	212

Figure 4.



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Discussion & Conclusions

- Implications
 - High recognition of some modes
 - Overidentification of ELMing modes
 - Confusion in L modes in QH modes

• Future research

- Expand database
- Include more features
- Refine hyperparameters

Questions?

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Picture credit:

https://www.google.com/url?sa=i&url=https%3A%2F%2Fwww.pngguru.com%2Ffree-transparent-background-png-clipart-emdt e&psig=AOvVaw3PqRBiJFXCOTDJTnXj9wqo&ust=1596661652244000&source=images&cd=vfe&ved=0CAMQjB1qFwoTCKjI k7i6qusCFQAAAAAAAAAAAAAAA

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Selection Process: Model with Topology Features

 Model and accuracy score

ExtraTreeClassifier() 0.5517241379310345 DecisionTreeClassifier() 0.6275862068965516 RadiusNeighborsClassifier() nan KNeighborsClassifier() 0.4206896551724138 ClassifierChain(base estimator=RandomForestClassifier()) nan OutputCodeClassifier(estimator=RandomForestClassifier()) 0.7241379310344828 OneVsOneClassifier(estimator=RandomForestClassifier()) 0.7206896551724138 OneVsRestClassifier(estimator=RandomForestClassifier()) 0.7344827586206897 SGDClassifier() 0.06206896551724138 RidgeClassifierCV(alphas=array([0.1, 1., 10.])) 0.46551724137931033 RidgeClassifier() 0.4913793103448276 PassiveAggressiveClassifier() 0.1172413793103448 AdaBoostClassifier() 0.18226600985221678 GradientBoostingClassifier() 0.603448275862069 BaggingClassifier() 0.6655172413793103 ExtraTreesClassifier() 0.7931034482758621 RandomForestClassifier() 0.7758620689655172 BernoulliNB() 0.23793103448275868 CalibratedClassifierCV() 0.45517241379310347 GaussianNB() 0.6482758620689655 LabelPropagation() 0.1586206896551724 LabelSpreading() 0.16896551724137931 LinearDiscriminantAnalysis() 0.4827586206896552 LinearSVC() 0.2 LogisticRegression() 0.3931034482758621 LogisticRegressionCV() 0.4525862068965517 MultinomialNB() nan NearestCentroid() 0.3275862068965517 NuSVC() 0.3448275862068966 Perceptron() 0.07931034482758621 QuadraticDiscriminantAnalysis() 0.013793103448275862 GaussianMixture() 0.0