

# Modeling of Necking Area Reduction of Carbon Steel in Hydrogen Environment Using Machine Learning Approach

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## Introduction

- Natural gas transport favors low carbon and alloy pipelines for their cost-effectiveness, weldability, longevity, and strength.
- Repurposing gas pipelines for hydrogen: Address regulatory, safety, and HE challenges.
- Addressing HE in Pipelines: Traditional mitigation strategies face adaptation challenges:



- Specialty Alloys:** Efficient, yet require costly specialized production.
- Protective Coatings:** Offer barriers, but come with expense and infrastructure modifications.

- Evaluating steel's embrittlement is resource-intensive; machine learning offers a more efficient, cost-effective solution.

### Objectives

- Utilize ML to analyze hydrogen behavior in low carbon and low-alloy steels under high pressures.
- Evaluate several ML techniques to best predict HE's impact on mechanical properties, especially reduction in area.
- Provide guidance on material selection for hydrogen pipeline construction based on steel behavior in hydrogen environments.

## Methodology

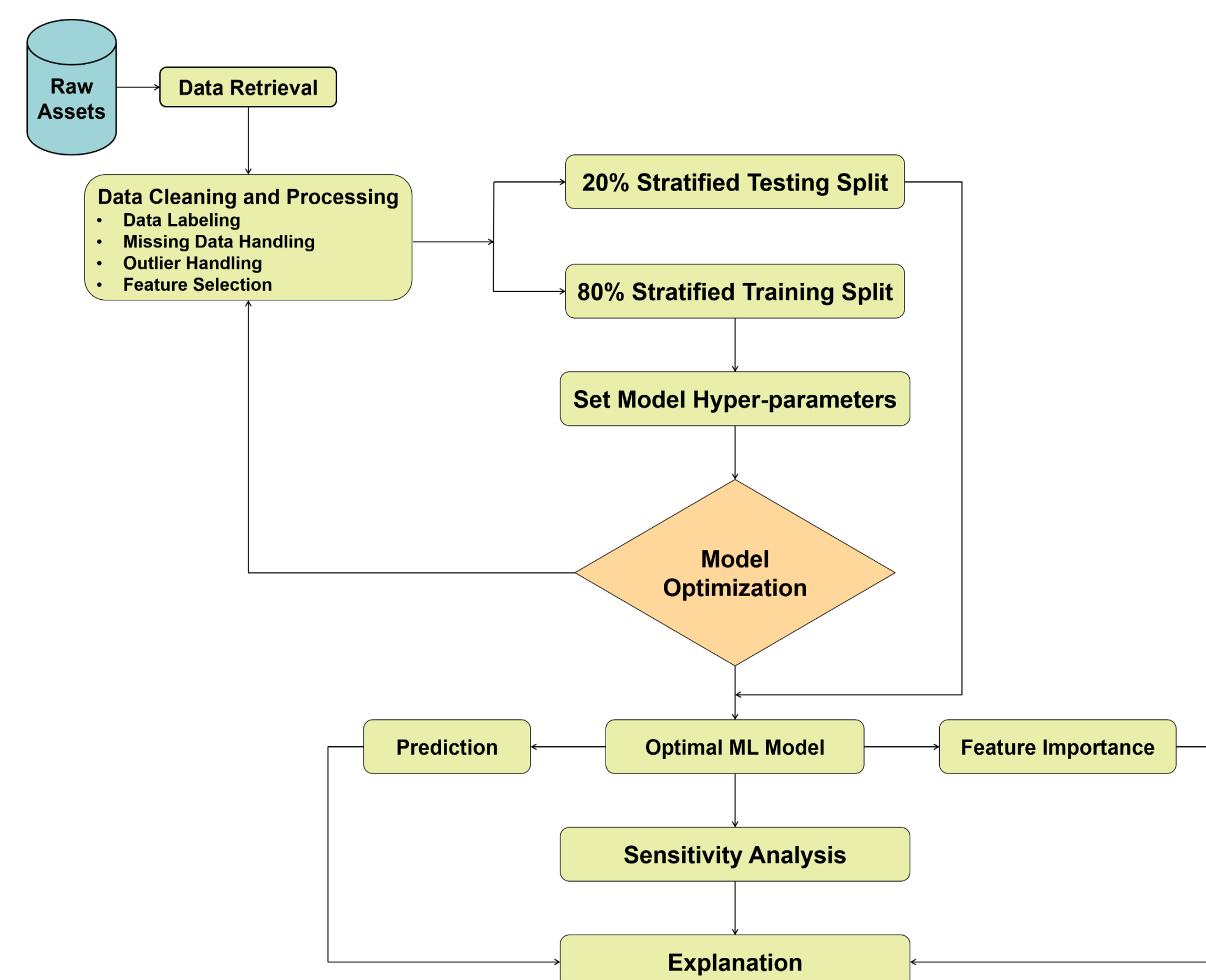
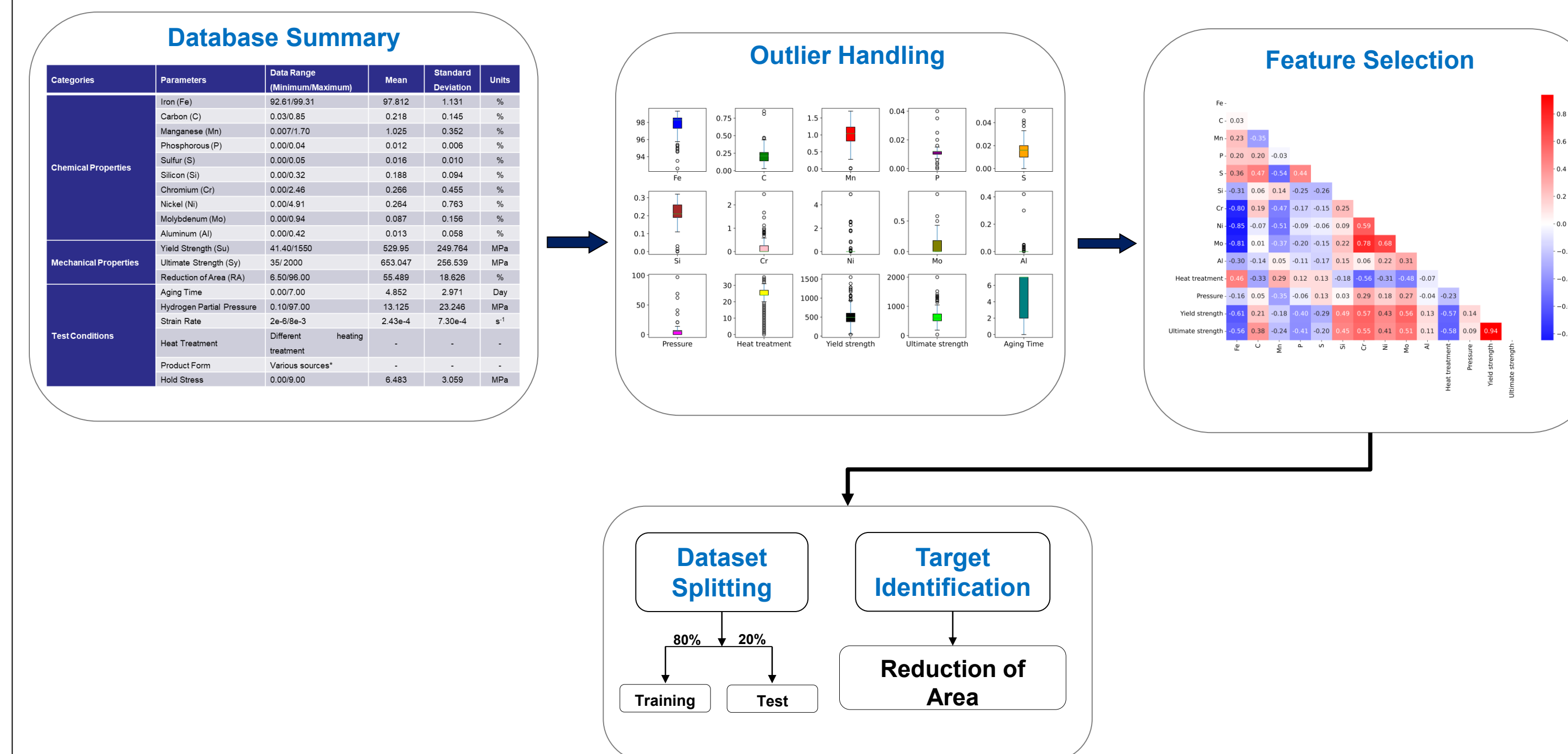


Fig. 1. Framework for Data Collection, Preparation, and Analysis.

## Results and Discussions

### Data Processing



### Machine Learning Model

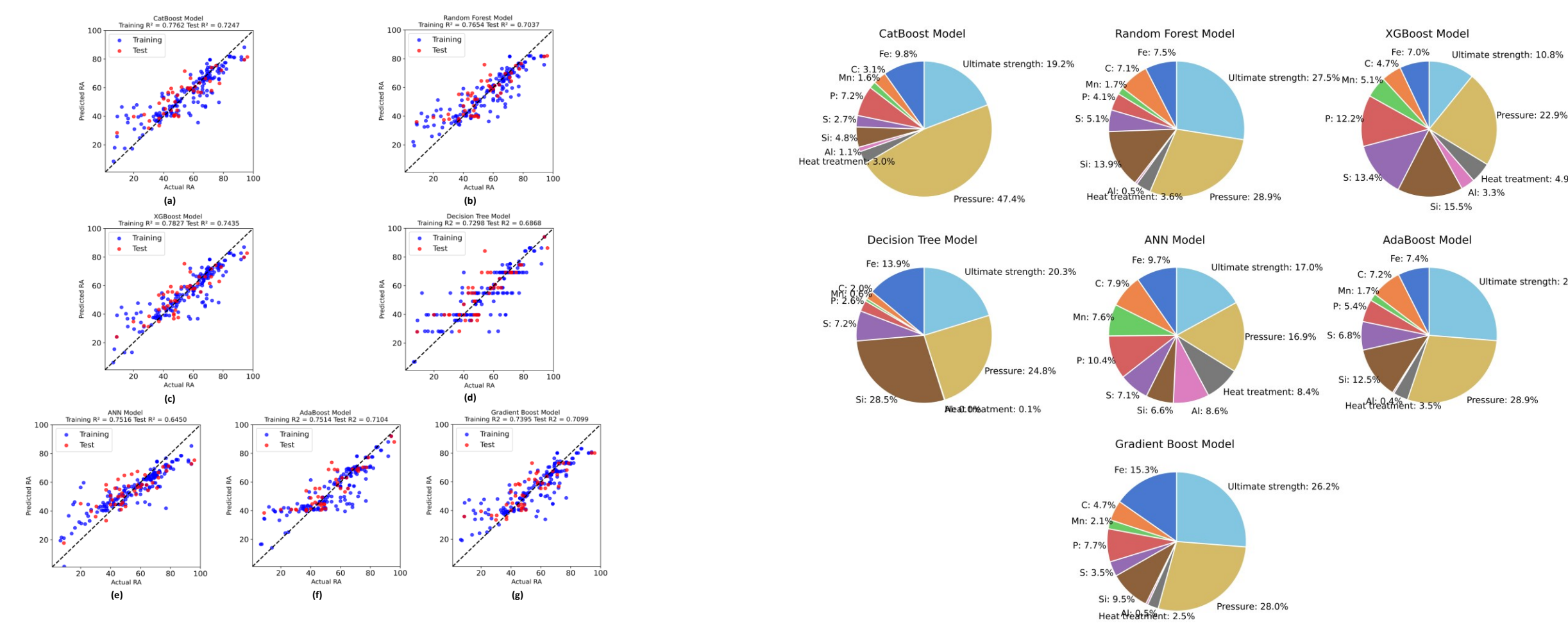


Fig. 2: Comparison of correlations between Actual RA and Predicted RA of seven ML models

Fig. 3: Comparison of different features in predicting reduction of area using seven machine learning models

### Performance Evaluation

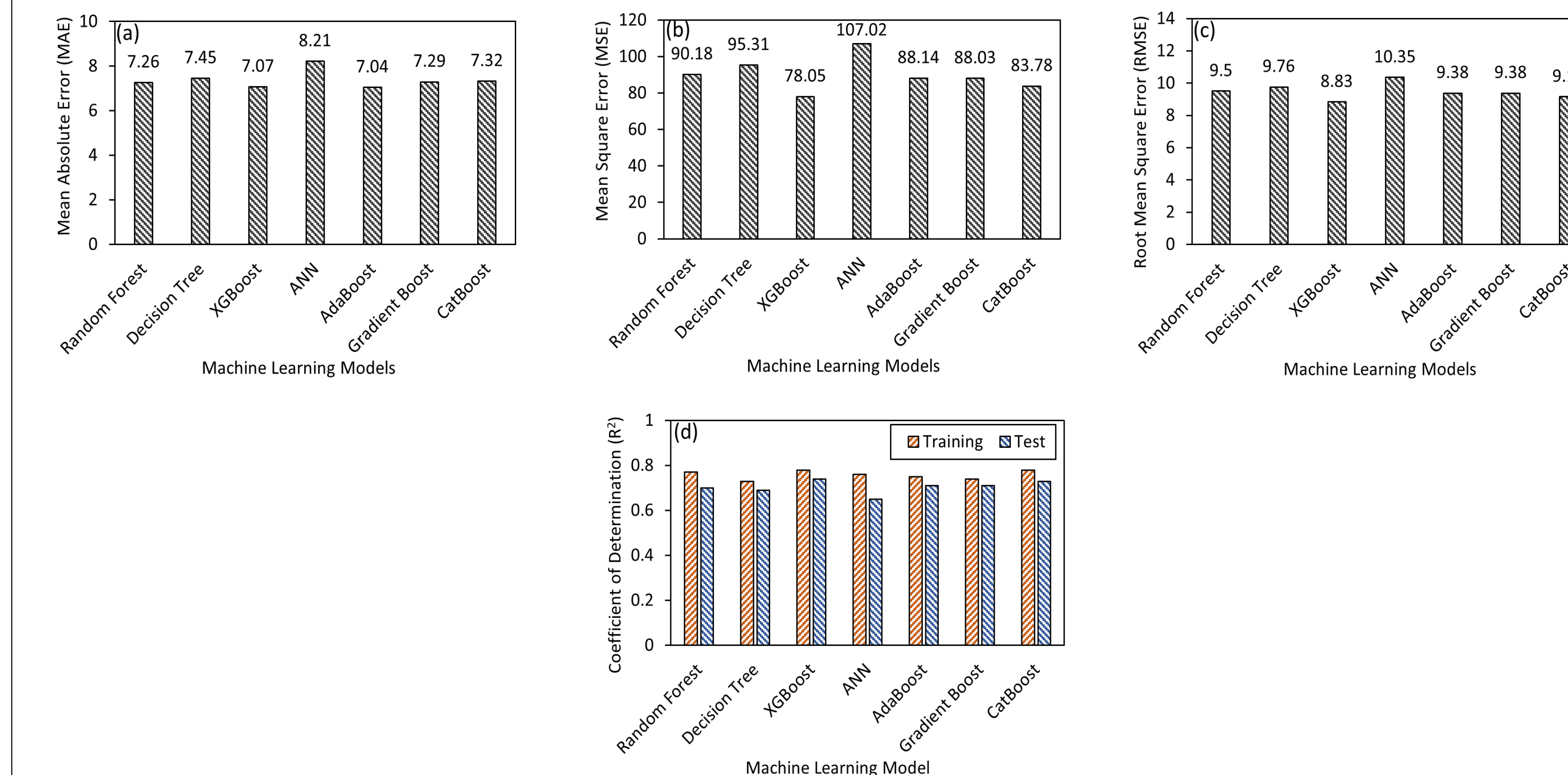
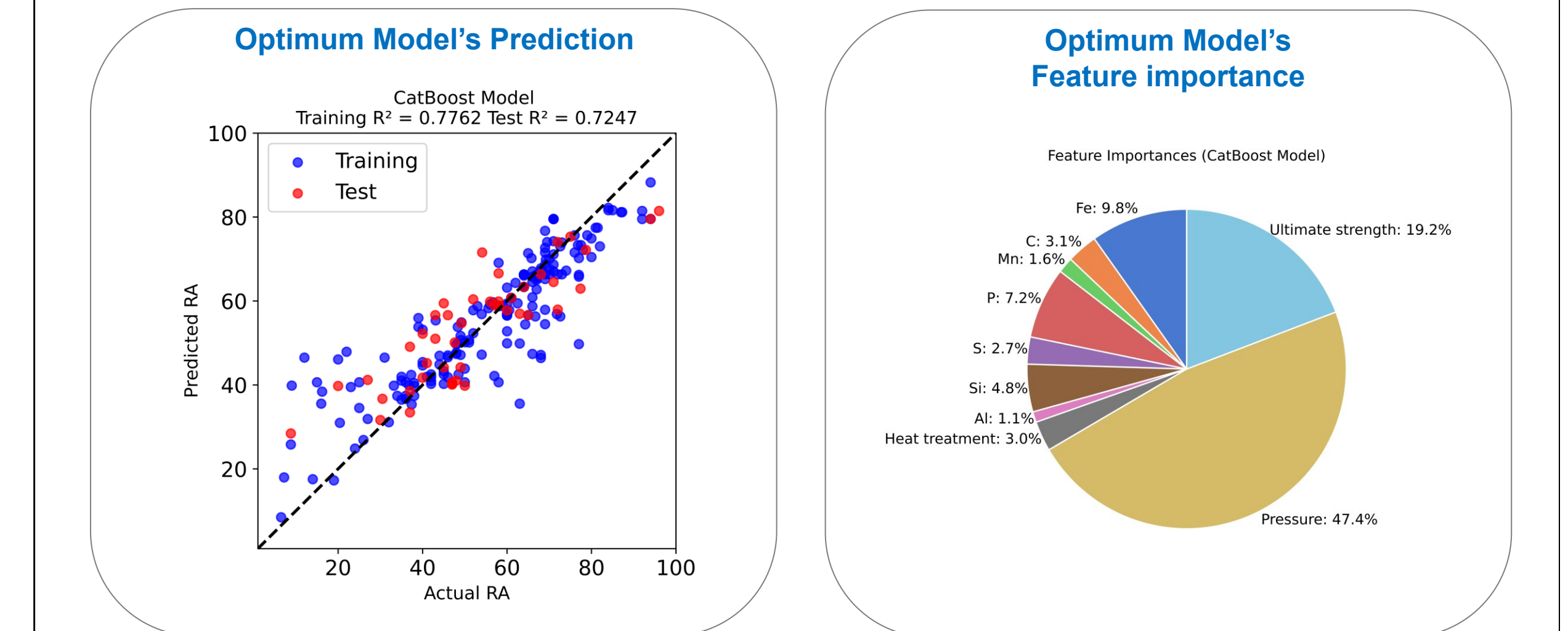


Fig. 4: Comparison of ML models performance based on (a) MAE, (b) MSE, (c) RMSE, and (d) R<sup>2</sup>

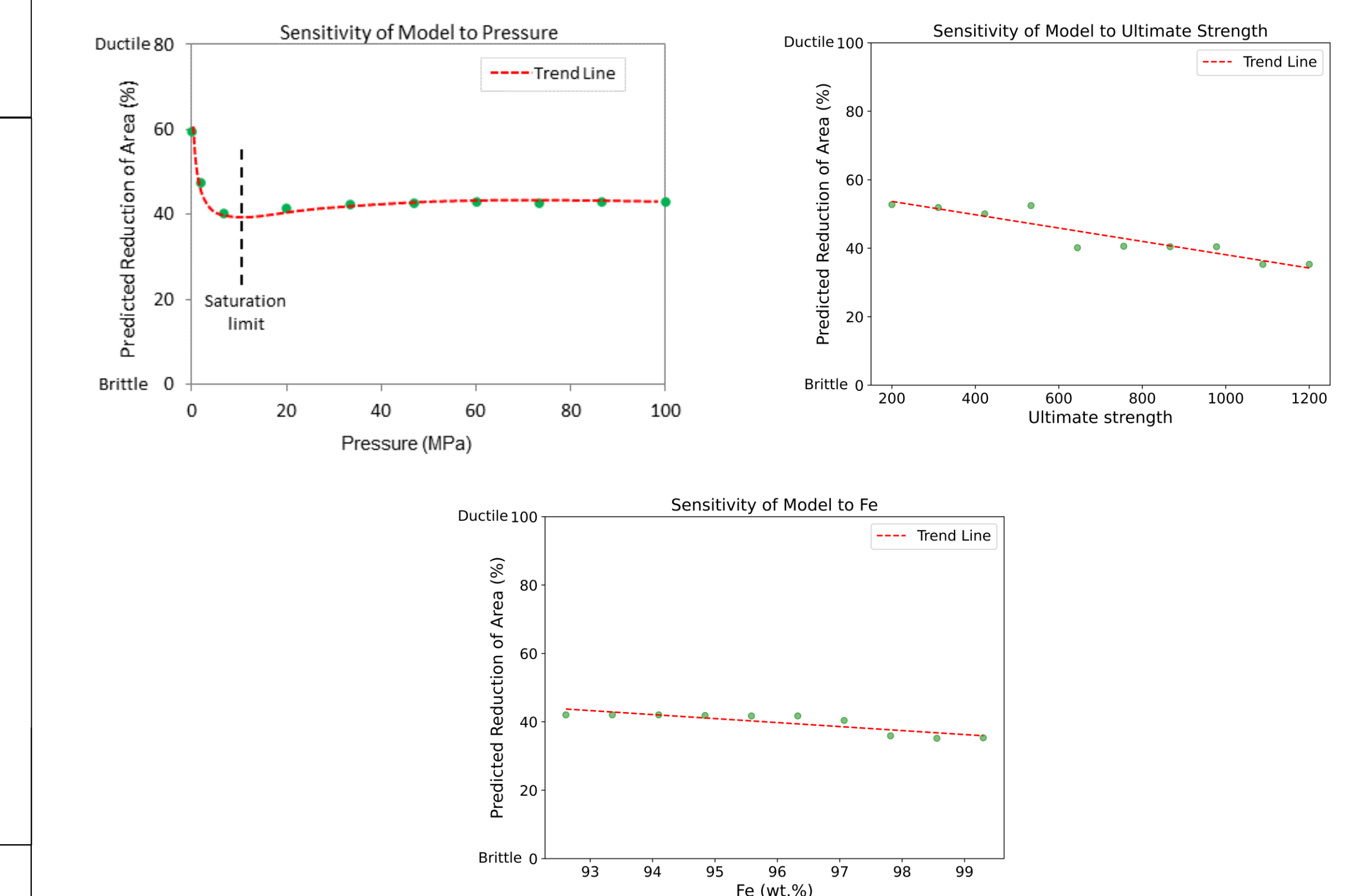
### Optimum Model Selection

Model Name	Coefficient of Determination	MSE	RMSE	MAE	Pressure	Su	Fe	C	Si	S	P	Al	Mn	HT*
RF	0.77	90.18	9.50	7.26	1	2	4	5	3	6	7	10	9	8
DT	0.73	89.31	9.76	7.45	1	2	4	7	8	9	3	10	5	6
XGBoost	0.78	78.05	8.83	7.07	1	5	6	9	2	3	4	10	7	8
ANN	0.75	88.04	9.39	8.18	1	2	4	7	10	9	3	5	8	6
AdaBoost	0.75	88.14	9.38	7.84	1	2	4	6	3	5	7	10	9	8
GB	0.74	88.03	9.38	7.29	1	2	3	6	4	7	5	10	9	8
CatBoost	0.78	83.78	9.15	7.32	1	2	3	6	5	8	4	10	9	7
Linear	0.76	90.07	9.48	7.38	1.0	2.4	4.0	6.6	5.0	6.7	4.7	9.3	8.0	7.3
Heat Treatment														

### CatBoost Model



### Model Sensitivity Analysis



## Conclusions

- CatBoost Model predicts hydrogen embrittlement in steel with MAE: 7.32, MSE: 83.78, RMSE: 9.15 (Training R<sup>2</sup>: 77.62%, Testing R<sup>2</sup>: 72.50%).
- Top Influencers: Pressure (47.4%) and UTS (19.2%) are critical, with elemental contributions from Iron, Phosphorus, and Silicon.
- Minor Elements: Mn, Al, and S have a combined, yet noteworthy impact.
- Application: Provides insights for material optimization, reducing hydrogen-induced risks.

## References

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