



# Online Transaction Fraud Detection

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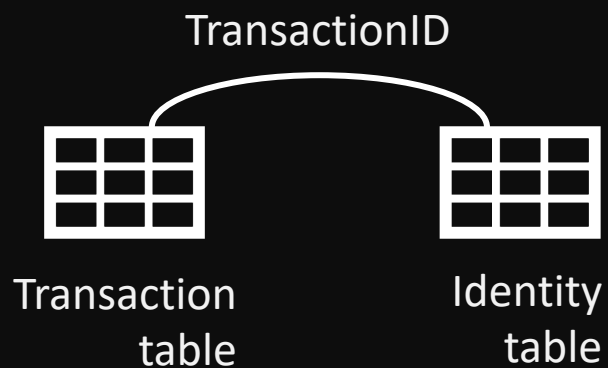
Jung-a Kim



# Online Payment Fraud

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- Online sales make up 17% of all global sales.
- For every \$1 of fraud from chargebacks, e-commerce businesses lose an extra \$2.94.
- Fraud also impacts brand and customer loyalty.

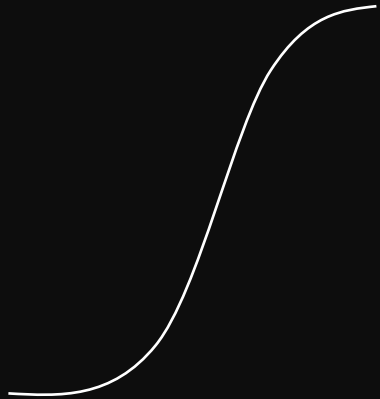


## Dataset

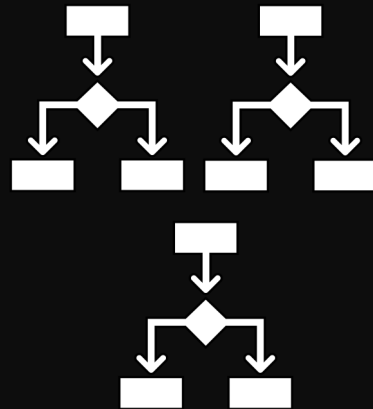
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- 590,540 transactions with labels
- 383 numeric features
  - e.g) Vesta- engineered rich features , time(D), payment amount, count(C), identity information
- 50 categorical features
  - e.g) Product/service code, address, card type, device type, identity information
- Class imbalance
  - 3.5% fraudulent transactions

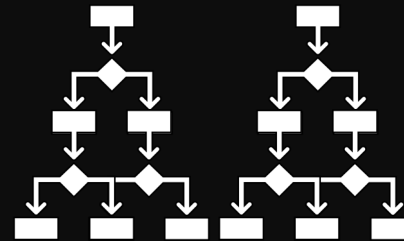
# Candidate models



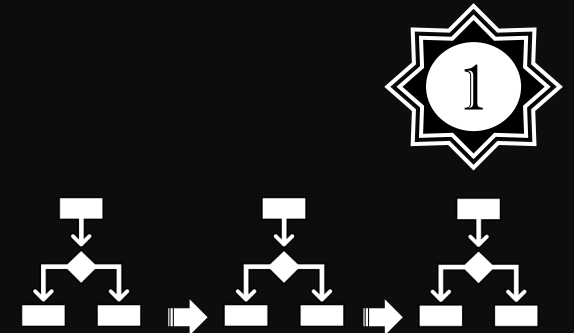
Logistic Regression



Random Forest

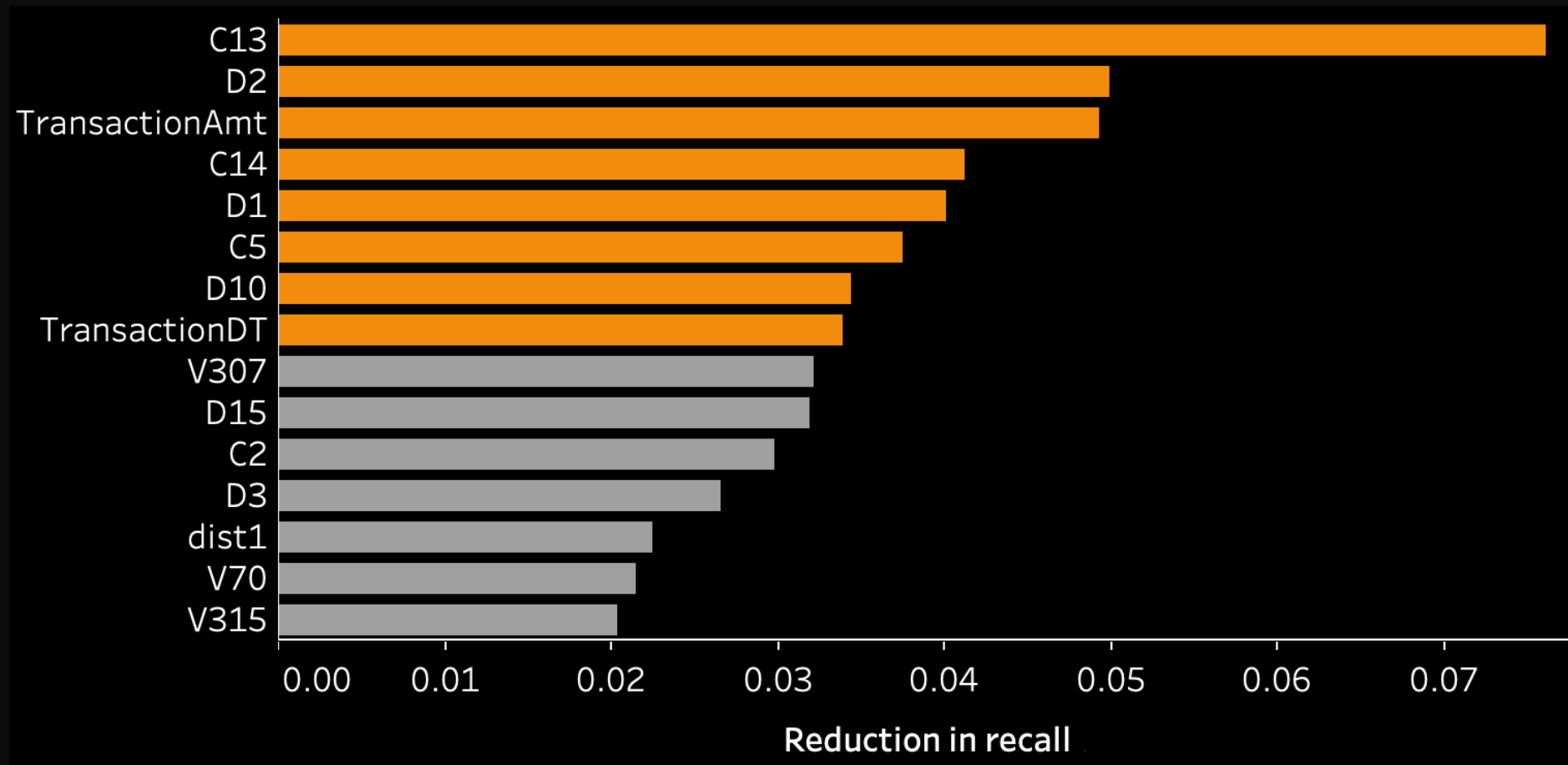


Extra Trees

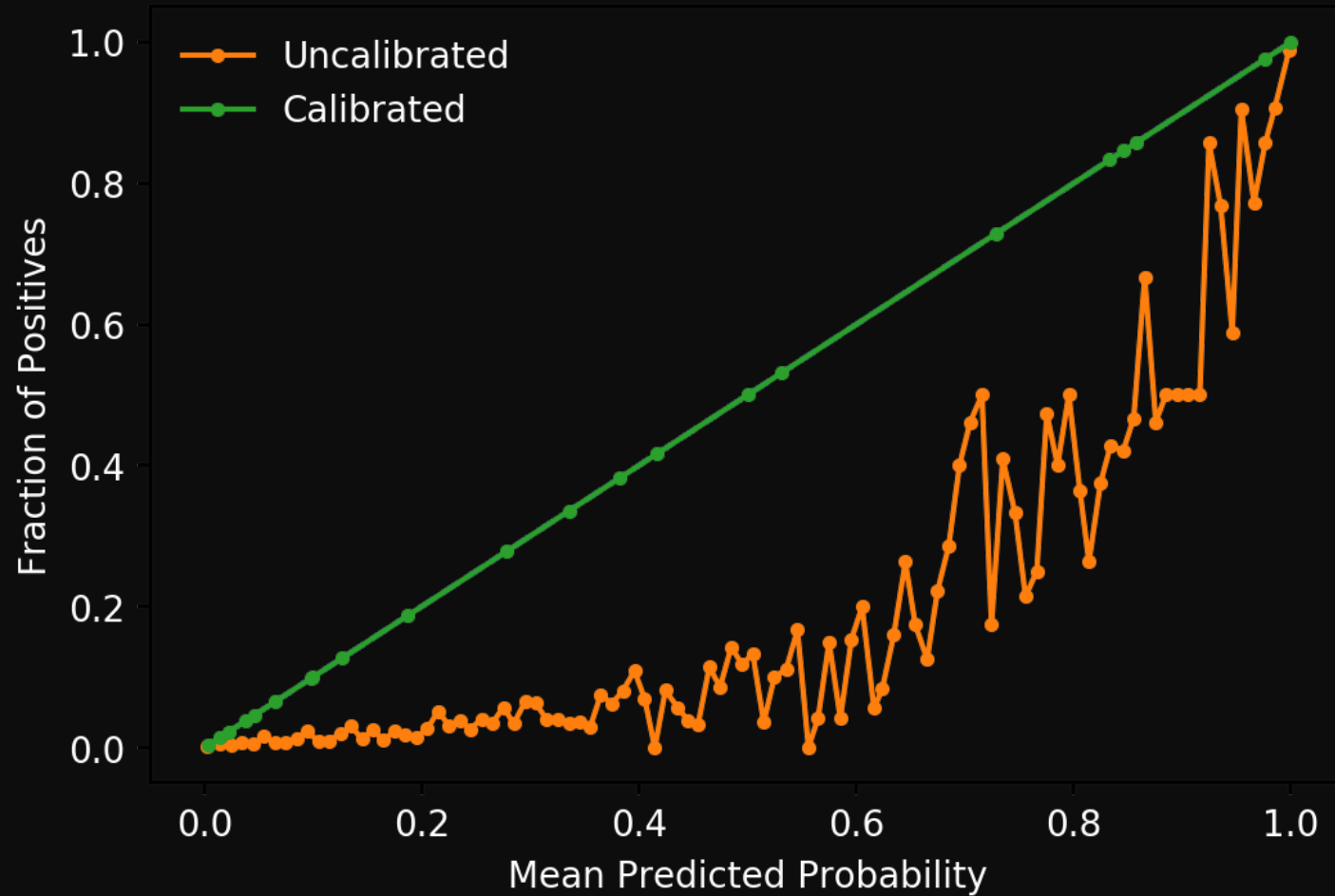


XGBoost

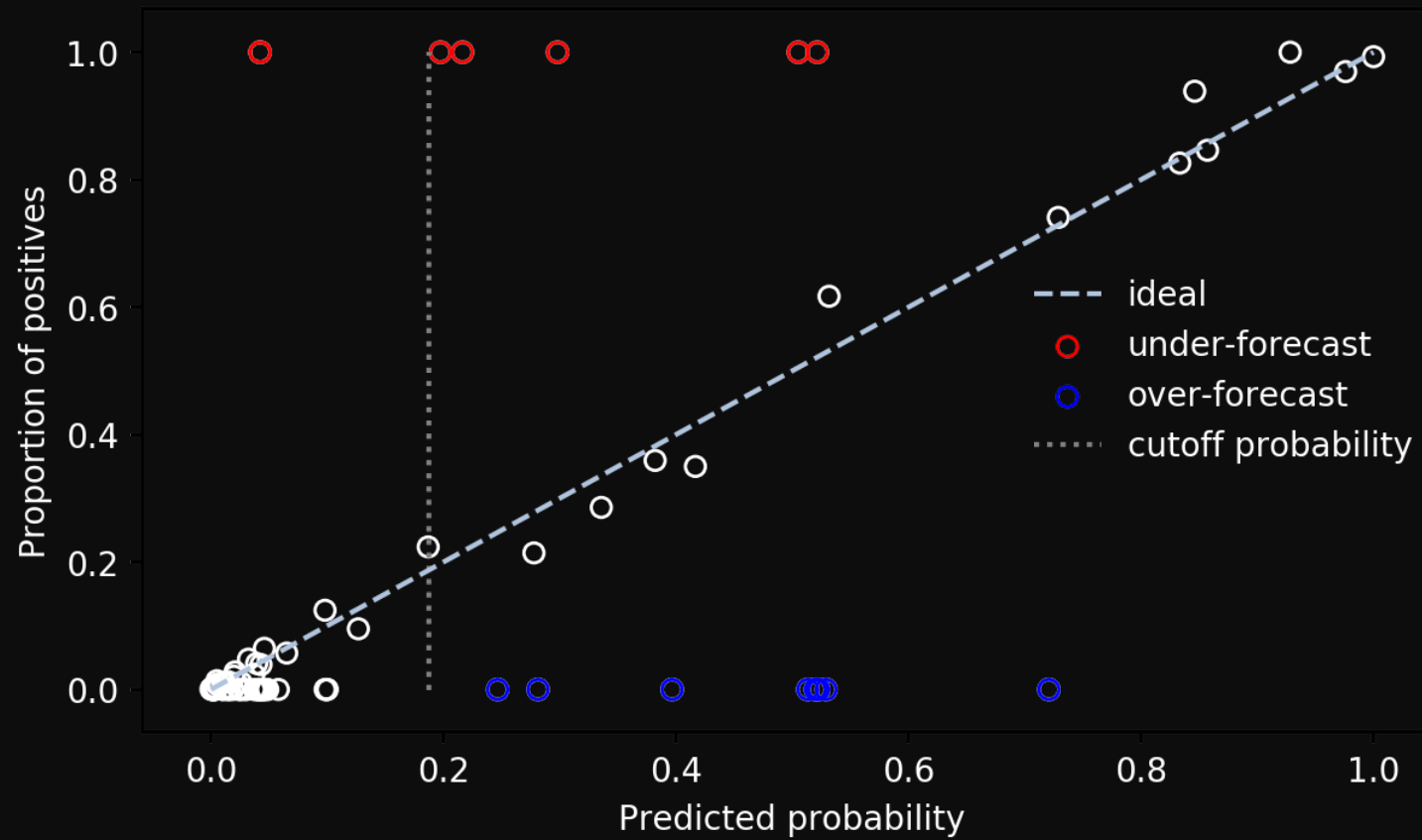
# Permutation Feature Importance



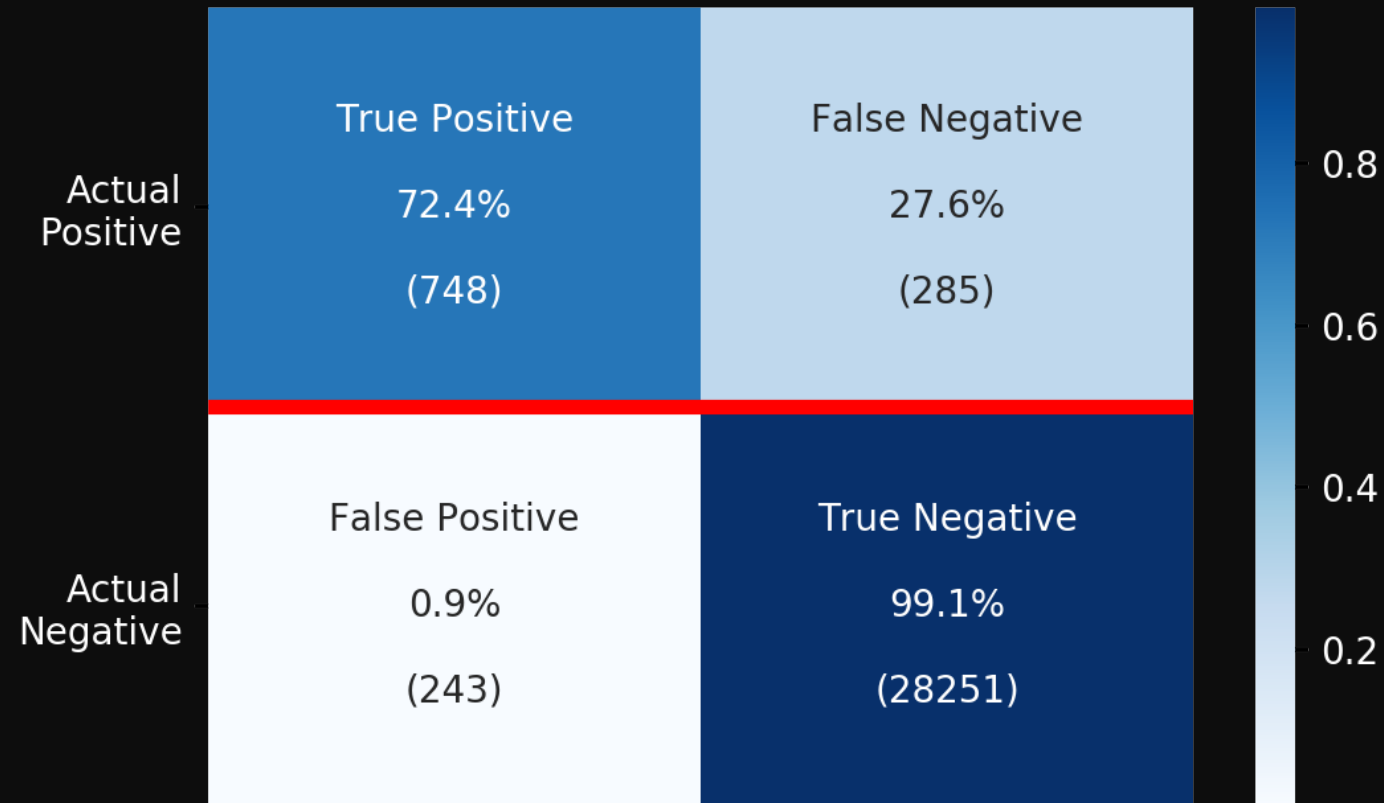
# Calibration on validation set



# Calibration on test set

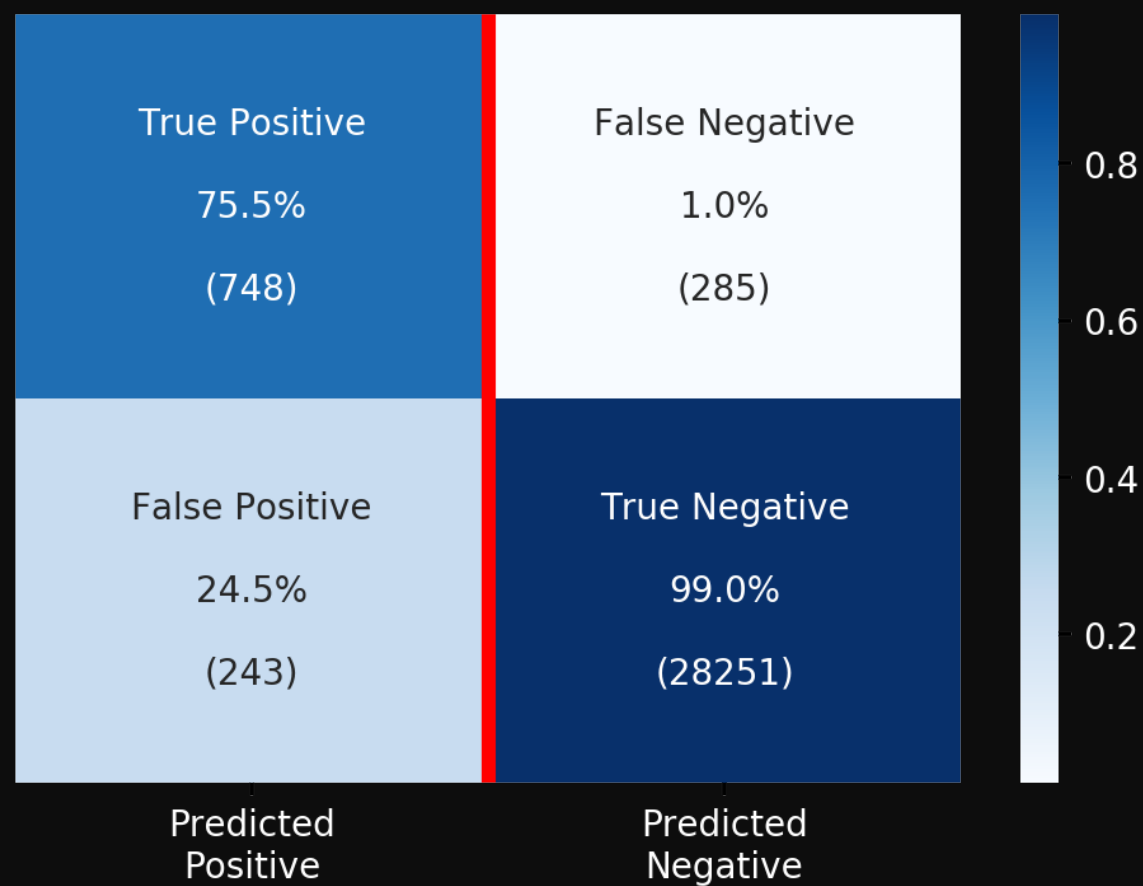


# Confusion Matrix





# Confusion Matrix



## Conclusion

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- Count, time, payment amount >>> product, address, card types, email domain
- Predicted probabilities  $\sim$  true probabilities
- Moderate recall and precision
- High specificity and negative predictive value

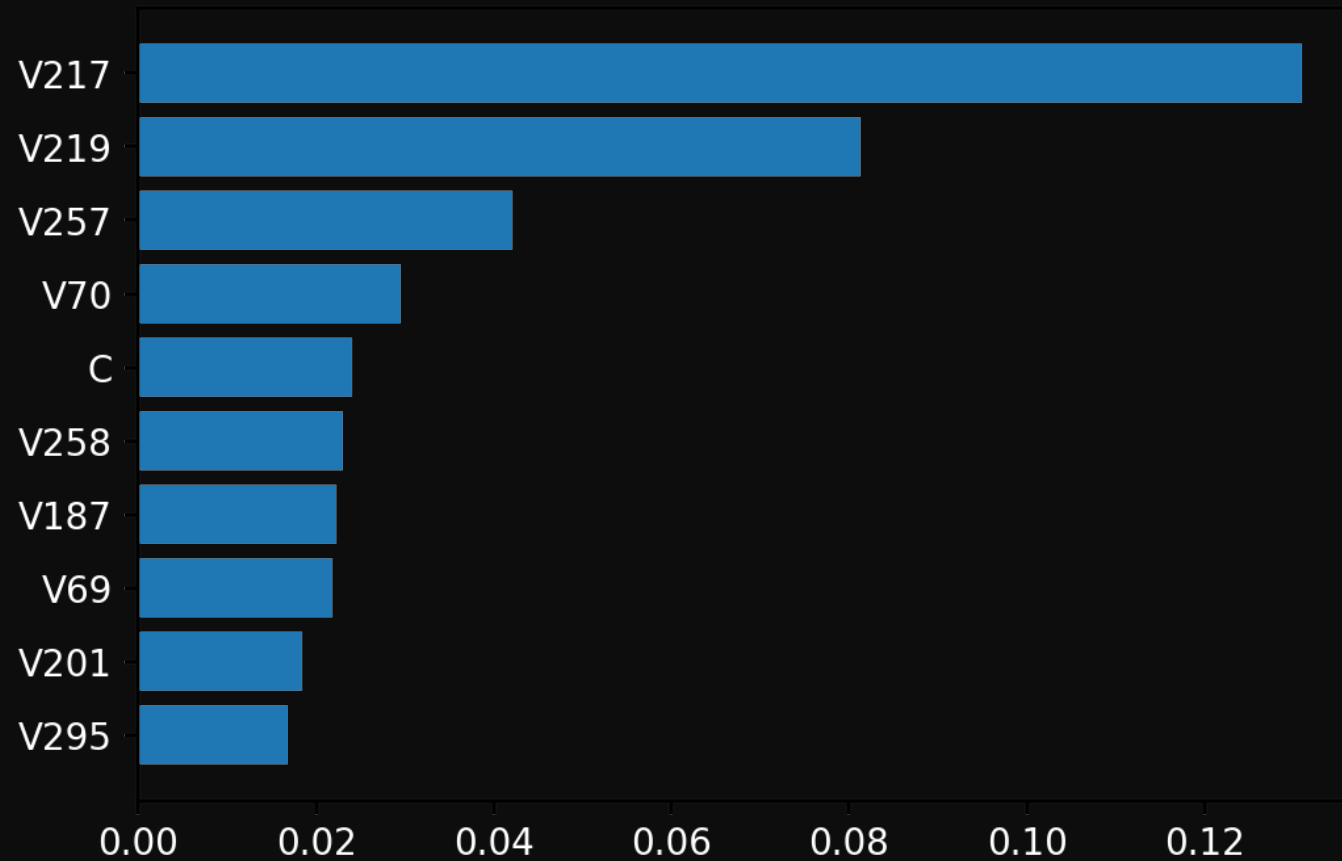
## Future work

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- Fine-tuning XGBoost
  - Tune Maximum depth, minimum samples-to-split, subsample ratio,  $\gamma$ ,  $\lambda$ , etc
- Simpler model
- Domain research

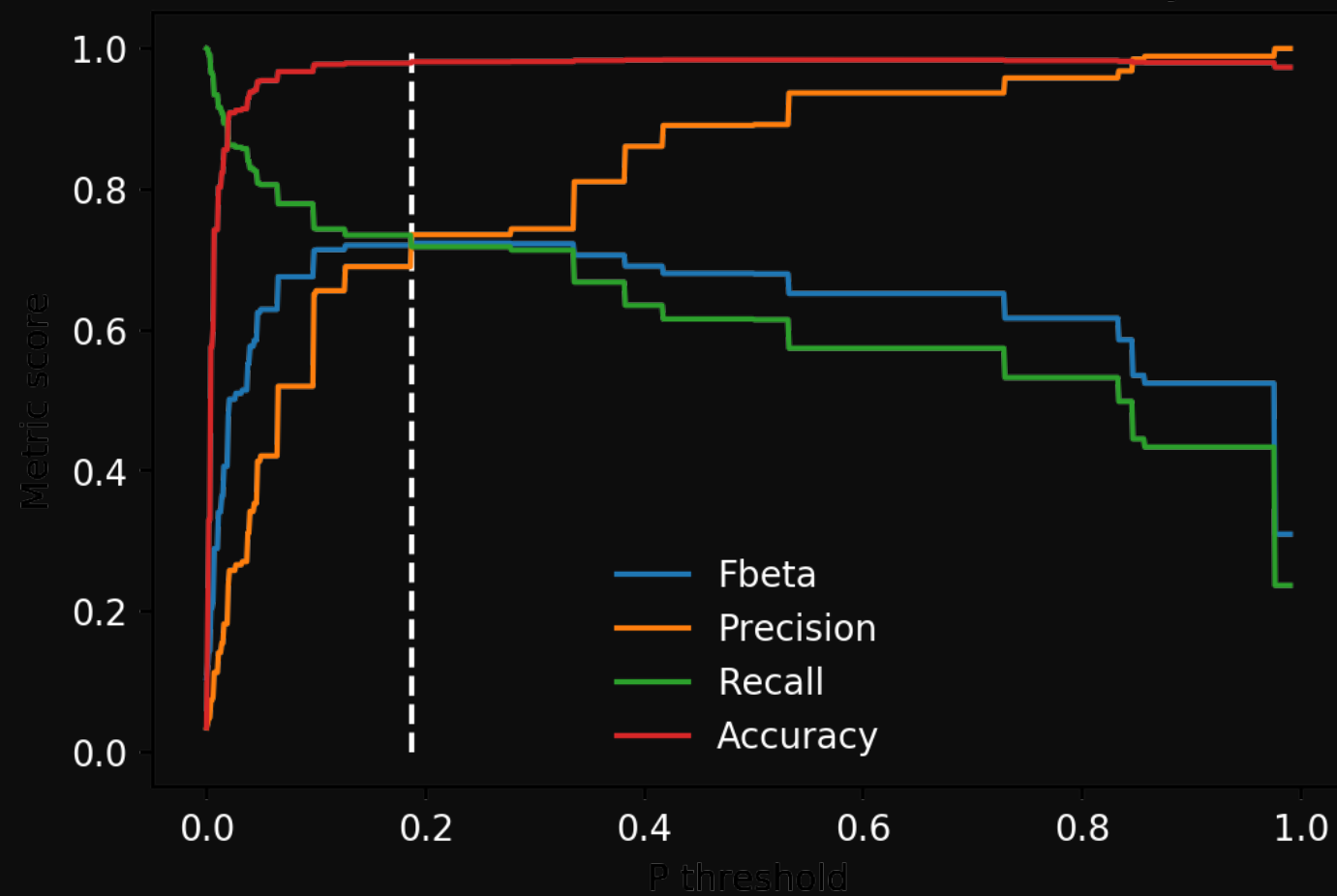
Thank you

# Feature Importance

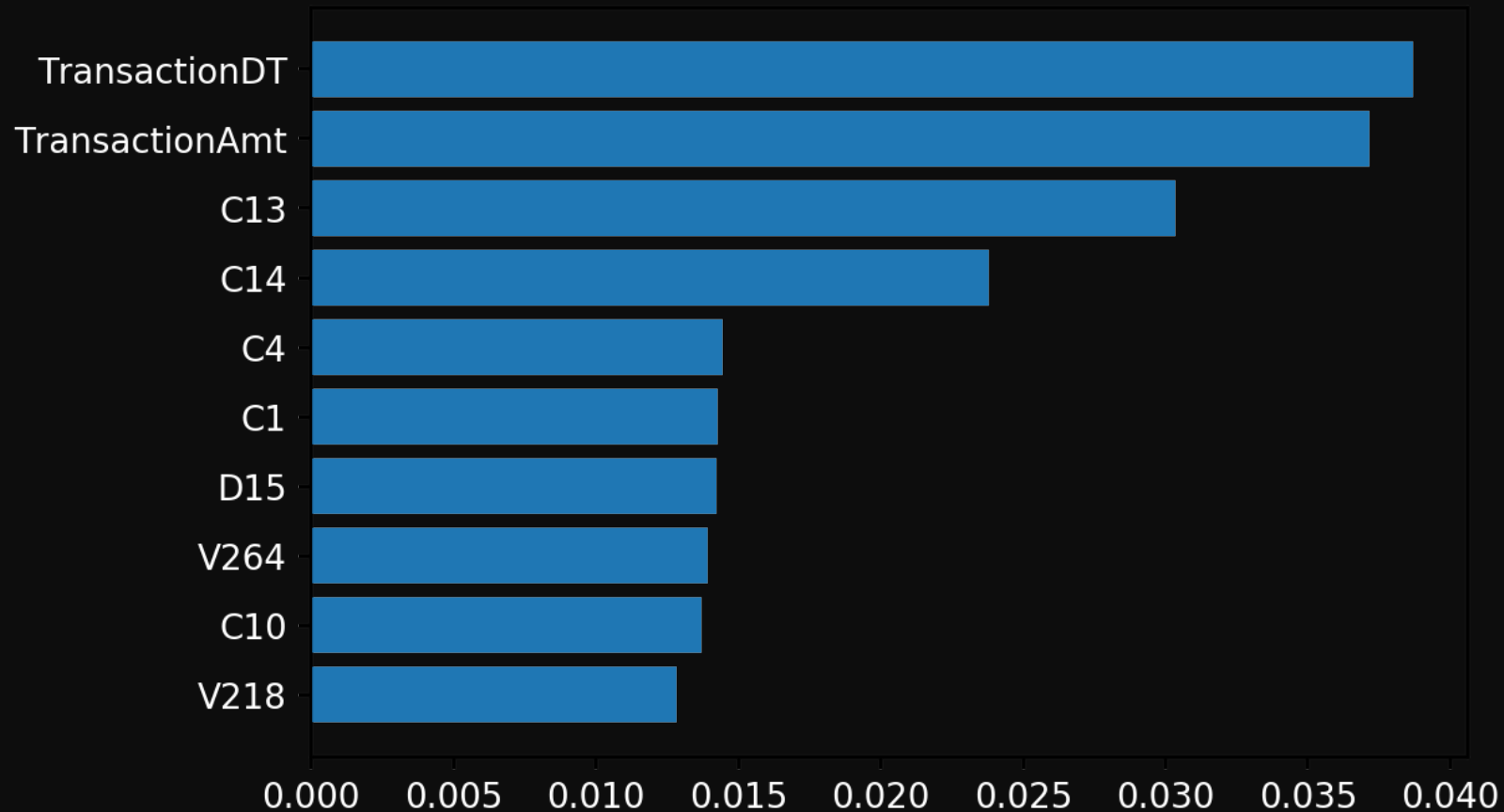


Vesta-engineered features  
are advantaged

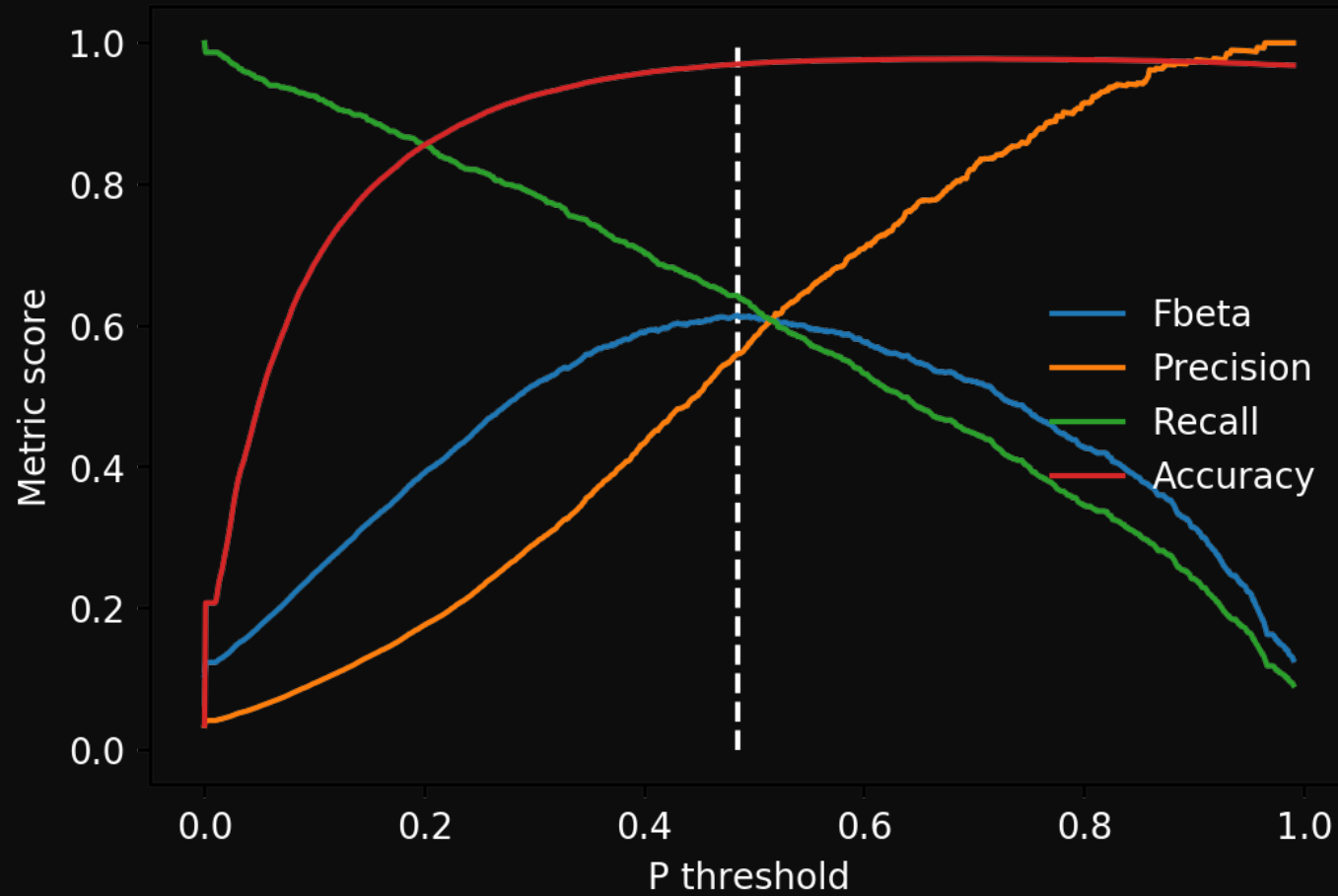
# Recall-Precision



# Random Forest Feature Importance



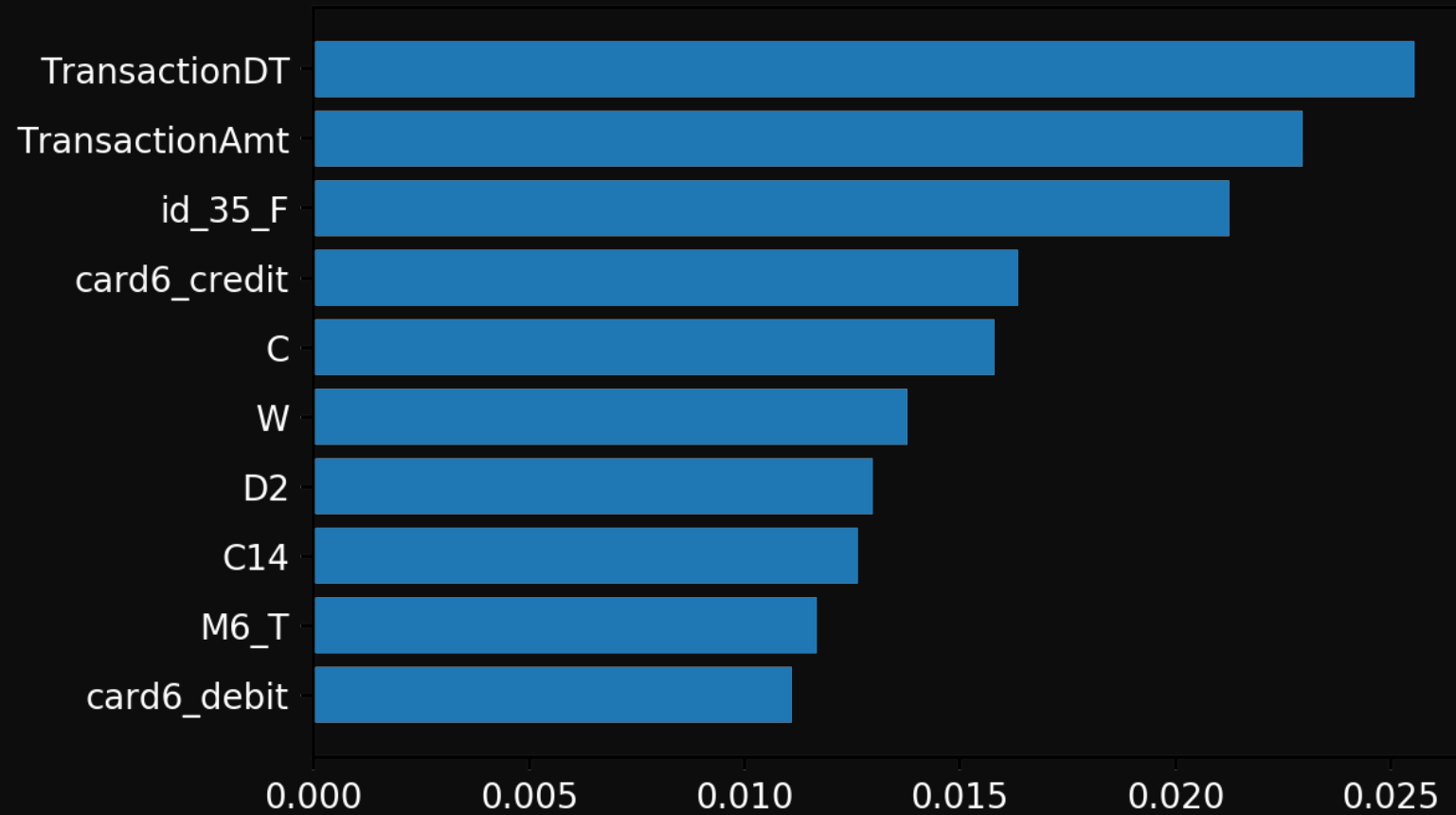
# Random Forest Recall-Precision





# Extra Trees Feature Importance

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# Extra Trees Recall-Precision

