

## Background and Significance



- Migraine affects >1 billion people globally and is a leading contributor to disability and reduced quality of life
- Despite its burden, migraine remains under-recognized and underdiagnosed in primary care settings

• Primary care providers often encounter nonspecific symptoms (e.g., fatigue, generalized pain), which may obscure migraine diagnosis

• Delayed or missed diagnosis may lead to increased emergency department utilization and progression to chronic migraine

• The widespread use of electronic medical records (EMRs) enables the use of routinely collected clinical and social data for earlier identification

• Predictive modeling approaches may help identify high-risk patients who could benefit from preventive care and timely intervention

## Aims

- Identify demographic, clinical, and social risk factors associated with migraine
- Evaluate the performance of machine learning and statistical models for detecting migraine risk using EMR-derived data

## Methods

• A cross-sectional analysis was conducted using de-identified EMR data from outpatient primary care encounters across Jefferson Health clinics.

### Study Population

- N = 59,088 adult outpatient visits
- Adult patients receiving routine primary care services
- Multi-site clinic representation including urban and suburban settings

### Predictor Variables

- Demographics Factors:** Age, Sex, Clinic Location
- Clinical factors:** Pain-related conditions, psychiatric comorbidities, vital signs
- Social risk indicators:** Health-related social needs (HRSN) documented in the EMR

### Model Development & Evaluation

- Two models evaluated:
  - L1-regularized logistic regression
  - Random Forest
- Data split into **70% training and 30% testing subsets**
- Model performance assessed using:
  - ROC-AUC and PR-AUC
  - Calibration and Brier score
  - Threshold-based diagnostic metrics

## Key Findings

### HRSN Status and Migraine Prevalence

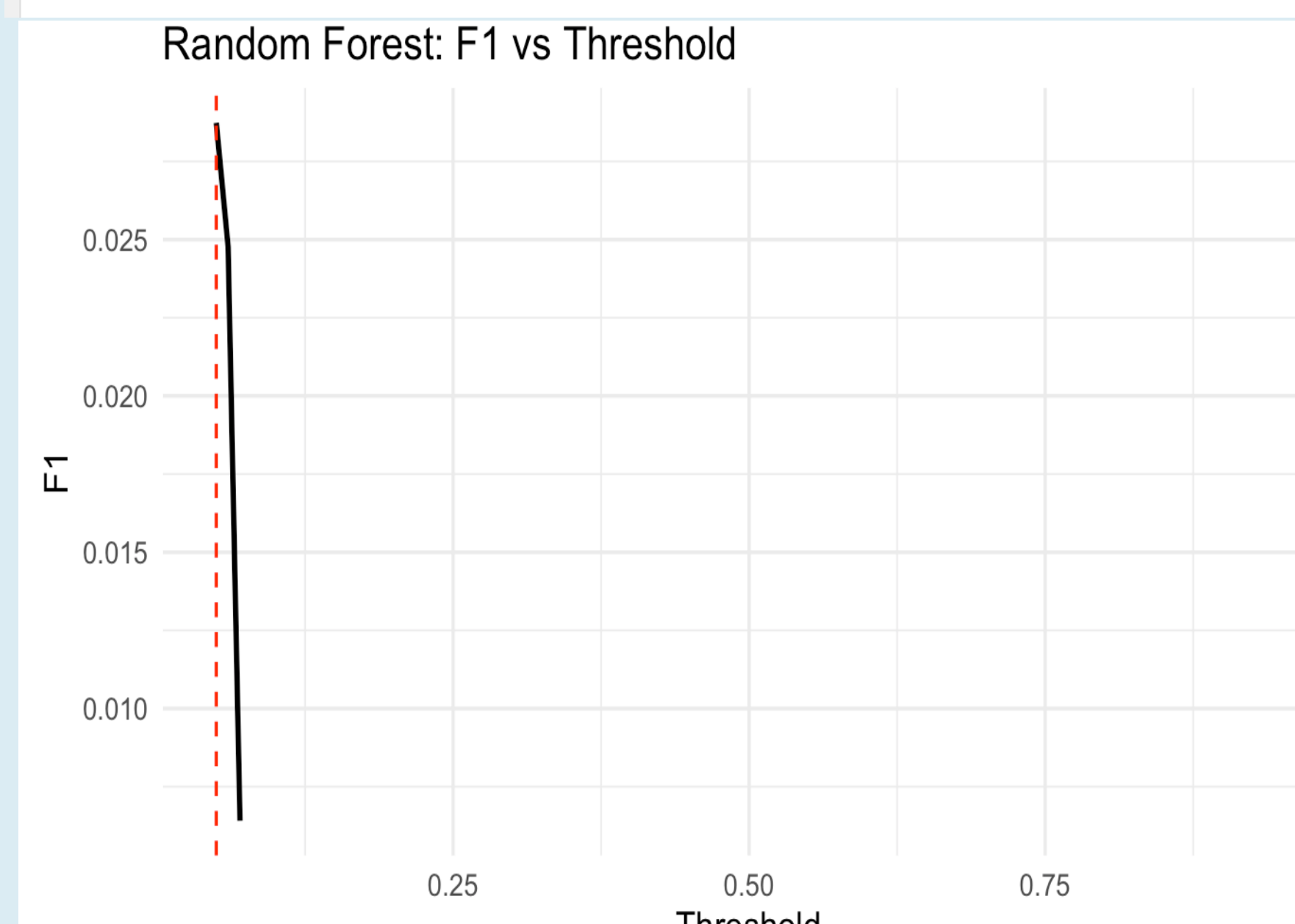
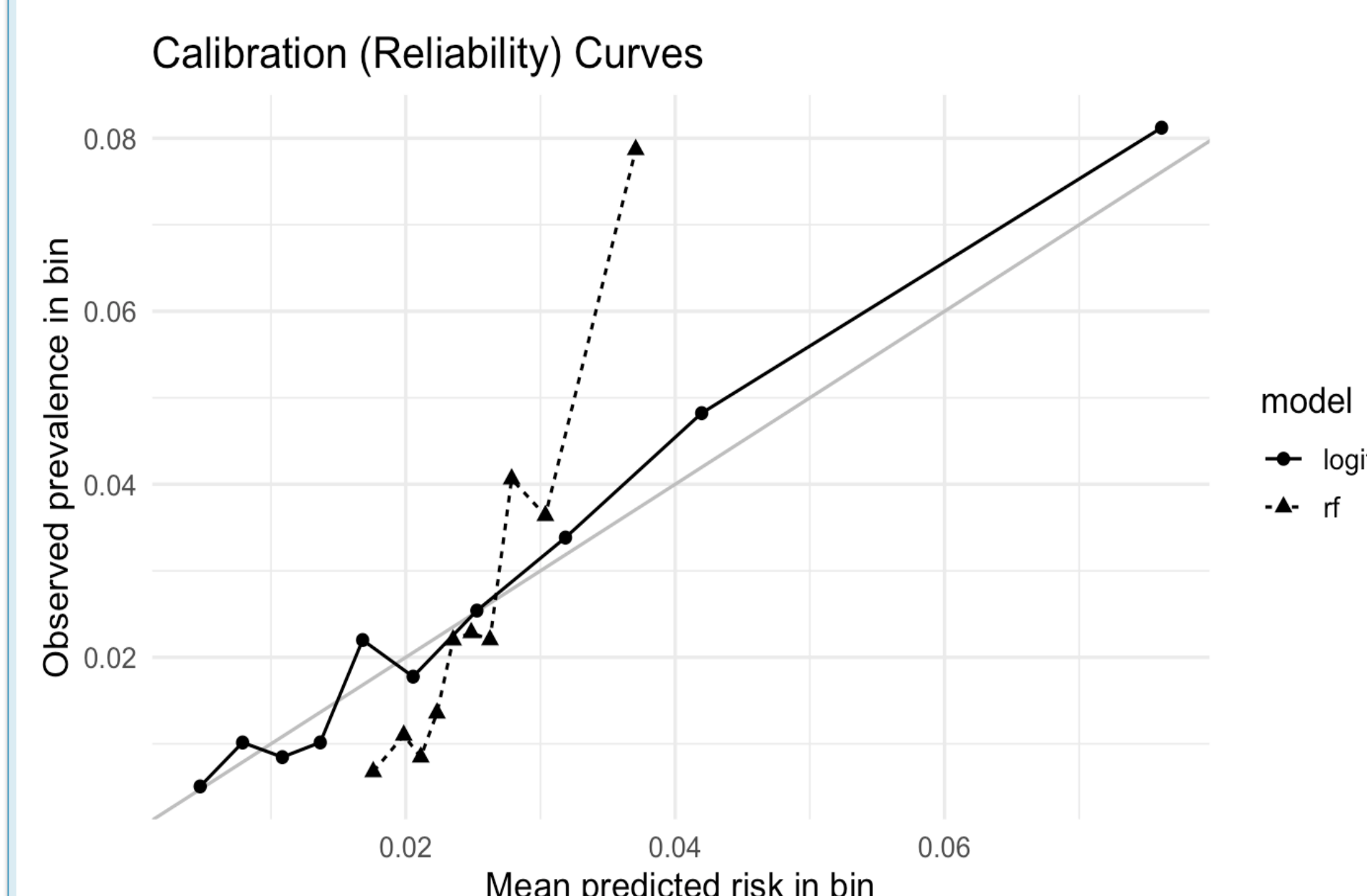
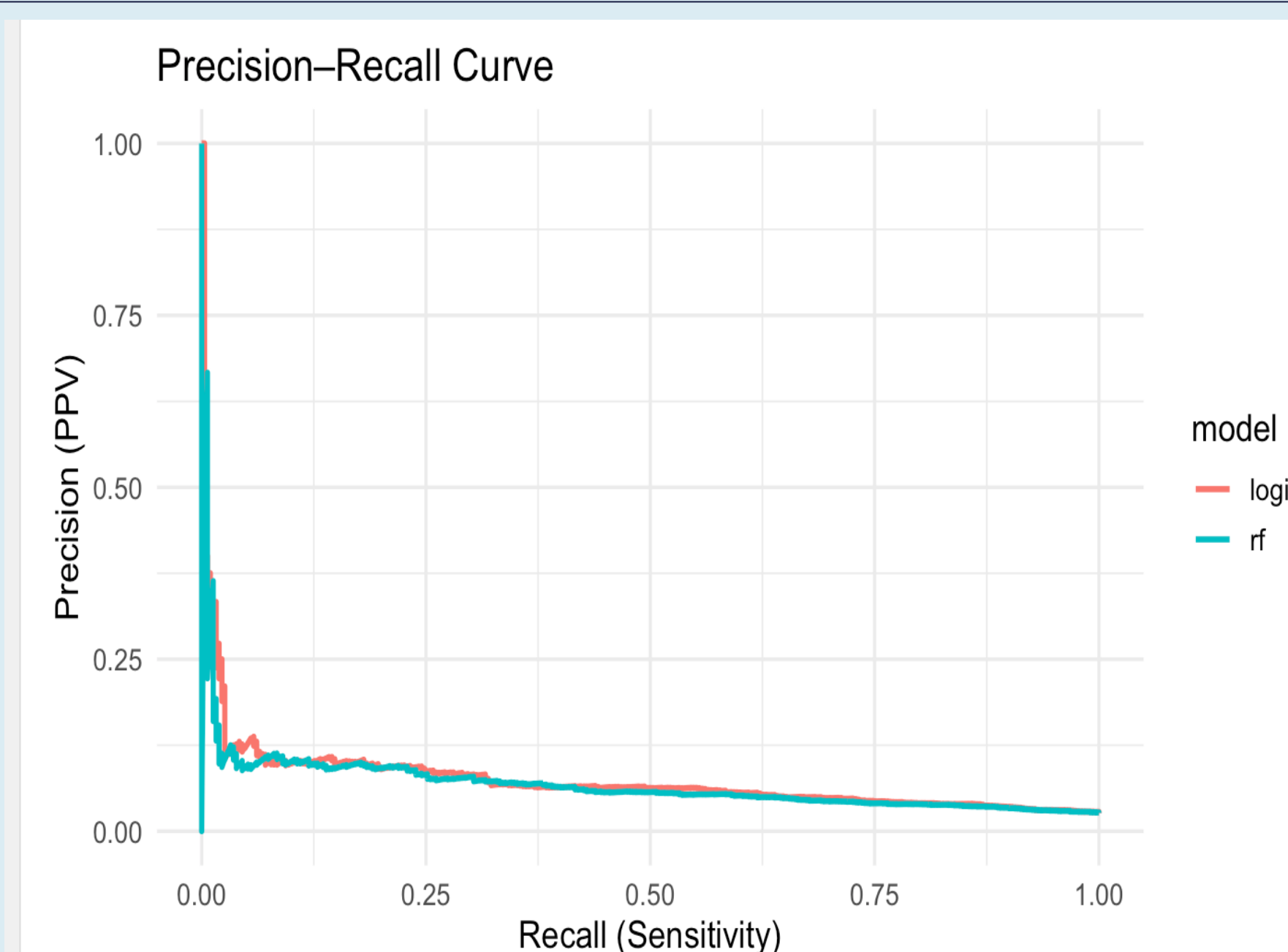
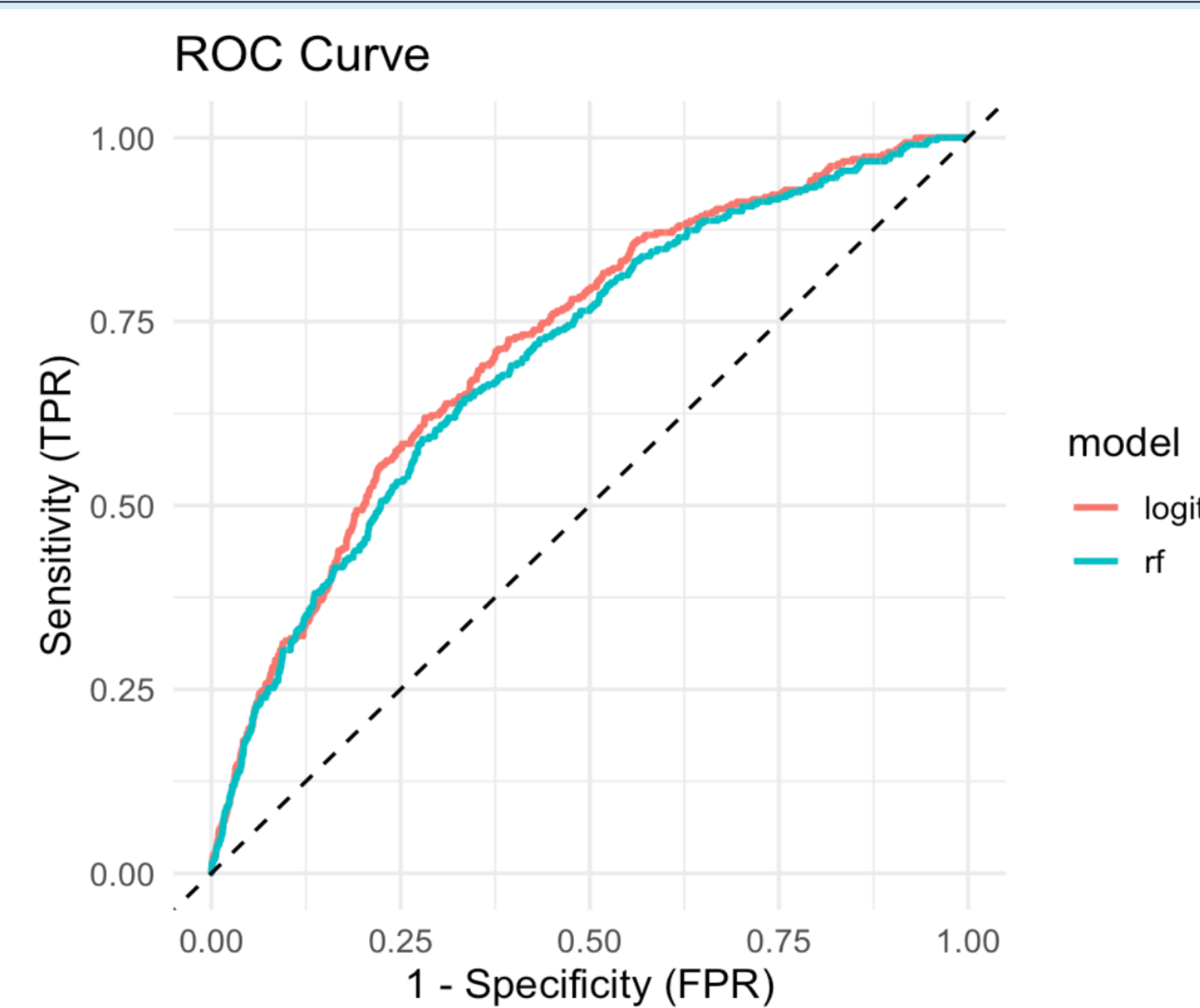
Any HRSN	Migraine Flag	N	%
No	No	53009	97.53
No	Yes	1345	2.47
Yes	No	4584	96.83
Yes	Yes	150	3.16

### Variable Importance Based on Impurity Index

Variable	Importance
Age	41.15
Sex - Male	27.09
Sex - Female	26.75
BMI	26.22
Asthma Dx Yes	15.19
Fibro Dx No	14.94
Depression Dx Yes	14.62
Anxiety Dx No	14.55

### Sensitivity Analyses: Model Performance on Test Set

Scenario	Model	ROC-AUC	PR-AUC	Brier
Same-Day Dx Only	Random Forest	0.736	0.083	0.024
Same-Day Dx Only	L1- Logistic	0.710	0.055	0.024
Exclude HRSN	L1- Logistic	0.720	0.073	0.025
Exclude HRSN	Random Forest	0.705	0.069	0.025
WHO BMI + ACC/AHA BP	Random Forest	0.745	0.101	0.024
WHO BMI + ACC/AHA BP	L1- Logistic	0.731	0.071	0.024



Threshold	ACC	Sensitivity	Specificity	PPV	NPV	F1	Youden
0.06	0.925	0.212	0.9439	0.0928	0.97803	0.1293	0.1568
0.05	0.891	0.303	0.9071	0.0808	0.97972	0.1276	0.2103

## Discussion

- Migraine prevalence was **2.53% (n = 1,495)**, suggesting underdiagnosis in primary care
- Higher prevalence observed among younger adults and female patients
- Strong associations with migraine were observed for psychiatric comorbidities, chronic pain conditions, and demographic factors such as age and sex
- Models showed moderate ROC-AUC (**0.705–0.745**) but low PR-AUC (**0.055–0.101**) due to rare outcomes
- Logistic regression showed better calibration and more reliable probabilities than Random Forest
- Clinical and social factors modestly improved predictive signal
- These findings support the growing recognition that migraine burden is influenced by complex biopsychosocial interactions
- EMR-based risk stratification tools may support conservative screening and guide referral to specialty care

## Limitations

- Cross-sectional design limits assessment of causality and temporal patterns
- ICD-10 coding may underestimate prevalence due to undiagnosed or undocumented cases
- Missing data (e.g., BP categories, HRSN) may have impacted model performance
- Limited clinical detail (e.g., headache symptoms, medications, referrals)
- Low prevalence (2.53%) led to class imbalance and reduced sensitivity
- Findings may have limited generalizability due to a single health system dataset

## Conclusion

- EMR-based predictive models can identify meaningful demographic, clinical, and social risk signals associated with migraine in primary care populations
- The low observed prevalence (2.53%) suggests substantial underdiagnosis in routine care
- Logistic regression provided more reliable and well-calibrated probability estimates for clinical use
- Models are best applied as **risk-flagging tools** to identify patients who may benefit from further evaluation and follow-up
- Future work should focus on improving sensitivity through enhanced data inputs and modeling approaches