

# Development of a Discrete-Time-Updating Algorithm to Predict Cannabis Use Disorder Among Arkansas Medical Marijuana Cardholders

**MSR68** 

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## **BACKGROUND**

- Currently, **24 states** have legalized cannabis for **adult recreational use** and **39 states** have legalized cannabis for **medical use**.<sup>1</sup>
- Among U.S. adults, the 2023 National Survey on Drug Use and Health (NSDUH) estimated that
  - ~58.9 million (22.9%) used cannabis in the past year.<sup>2</sup>
  - ~17.9 million (7.0%) met DSM-V criteria for CUD in the past year.<sup>2</sup>
- No CUD risk prediction tool has been developed among persons authorized to purchase medical cannabis
  - The underlying relationship between CUD risk factors and CUD development may look different among cannabis users relative to the general population<sup>3</sup>

Objective: To develop a discrete-time-updating algorithm to predict CUD risk within the next 90 days among Arkansas medical marijuana (MMJ) cardholders without a history of CUD.

## **METHODS**

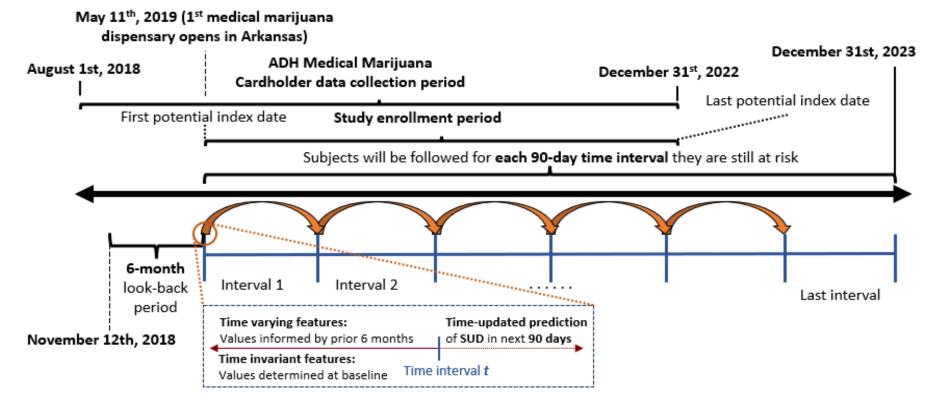
#### **Data Source**

• This study utilized statewide health insurance claims data linked to Arkansas Department of Health (ADH) MMJ cardholder data between November 2018 – December 2023 from the **Arkansas All-Payer Claims Database (AR-APCD).**<sup>4</sup>

#### **Study Sample**

- Index Date: May 11th, 2019 (Date 1<sup>st</sup> Arkansas MMJ dispensary opened) or receipt date of MMJ eligibility card, whichever came last
- Inclusion Criteria:
- i. MMJ cardholder eligibility between August 1<sup>st</sup>, 2018 and December 31<sup>st</sup>, 2022.
- ii. ≥ 18 years old at the time of the index date.
- iii. ≥ 180 days of continuous medical and pharmacy benefits enrollment prior to the index date
- iv. ≥ 180 days without a CUD diagnosis prior to the index date.
- **Follow-up**: Index date until whichever of the following first occurs: 1) New CUD diagnosis, 2) reach study end date (Dec. 31<sup>st</sup>, 2023), 3) health plan disenrollment, 4) death from any cause
- Data structure (Discrete-time-updating): Person-period dataset (subject follow-up split into 90-day time intervals), where CUD prediction for each time interval informed by prior 6 months of features

#### **Study Design**



#### Feature Space [n=174]

- Labeled **prognostic** if evidenced by prior literature [n=43], labeled **agnostic** otherwise [n=131]
- Demographics (Age, sex, insurance payer type) [n=3]
- Acute/chronic Comorbidities (Categorized using Clinical Classifications Software Refined) [n=114]
- Prescription Characteristics (Categorized using First Databank therapeutic classes) [n=52]
- Healthcare Utilization (E.g. Cumulative healthcare costs, hospitalization count) [n=5]

#### **Model Training/Testing**

- Train/test split: Random 70:30 split at person level
- Data balancing: 1:1, 1:3, and 1:10 random undersampling (RUS) of the majority class
- Hyperparameter tuning: 90 iterations with 5-fold cross validation

1. Random Forest-derived feature importance scores

- Feature selection: Two-pronged recursive feature elimination approach:
  - 2. Cox proportional hazards-derived p-values

# **METHODS**

**Classifiers:** Random Survival Forest (RSF), Gradient Boosting (GB), Support Vector Machine (SVM), Cox Proportional Hazards Survival Model (CPH), Random Forest (RF), Logistic Regression (LR).

**Performance Metrics:** Cumulative sensitivity/dynamic specificity area under the receiver-operating characteristic (C/D AUC), Brier Score, Sensitivity, Specificity, Positive Predictive Value (PPV), Negative Predictive Value (NPV)

### **RESULTS**

A total of **54,422 Arkansas MMJ cardholders** met eligibility, of which **1,479 (2.72%)** received a new CUD diagnosis during the follow-up period.

# Model Performance for the Prediction of Cannabis Use Disorder within the next 90 days among Arkansas Medical Marijuana Cardholders

	RUS 1:1		RUS 1:3		RUS 1:10	
Classifier	Mean C/D AUC	Mean Brier Score	Mean C/D AUC	Mean Brier Score	Mean C/D AUC	Mean Brier Score
Random Survival Forest <sup>a</sup>	0.7183	0.0237	0.7309	0.0082	0.7323	0.0054
Gradient Boosting <sup>a</sup>	0.7712	0.0250	0.7570	0.0079	0.7463	0.0054
Support Vector Machine <sup>a</sup>	0.7776	0.0242	0.7905	0.0075	0.7976	0.0048
Cox Proportional Hazards <sup>a</sup>	0.7683	0.0263	0.7802	0.0088	0.7888	0.0053
Random Forest <sup>b</sup>	0.6596	0.0102	0.6834	0.0061	0.6877	0.0053
Logistic Regression <sup>b</sup>	0.7045	0.0111	0.7721	0.0062	0.8068	0.0053
Random Forest <sup>c</sup>	0.6684	0.0098	0.6733	0.0061	0.6768	0.0053
Logistic Regression <sup>c</sup>	0.6790	0.0119	0.7118	0.0066	0.7233	0.0053

C/D AUC = Cumulative sensitivity/dynamic specificity area under the receiver-operating characteristic, NPV = Negative Predictive Value, PPV = Positive Predictive Value, RUS = Random Undersampling

 $<sup>{}^{\</sup>mathtt{c}}\mathsf{Time}$  interval excluded from model training

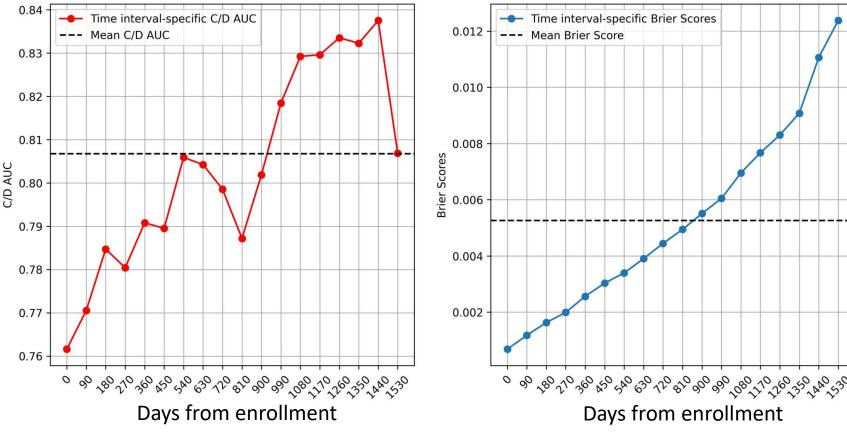
Top 10 Predictors of the Best Performing Models								
Logistic Regression (RUS 1:10)			Support Vector Machine (RUS 1:10)					
Rank	<b>Feature</b> <sup>a</sup>	Feature <sup>a</sup> χ <sup>2</sup> Value Featu		Coefficient Absolute Value				
1	Time Interval	0.085	Age	0.130				
2	Sex	0.064	Exposure/screening with infectious disease <sup>c</sup>	0.097				
3	Age	0.059	Medical examination/evaluation <sup>c</sup>	0.086				
4	Depression <sup>c</sup>	0.026	Back Pain <sup>c</sup>	0.083				
5	Tobacco Use Disorder <sup>c</sup>	0.024	Other perinatal conditions <sup>b</sup>	0.081				
6	Bipolar and Related Disorders <sup>c</sup>	0.020	Count of PCP visits <sup>d</sup>	0.075				
7	Malaise and Fatigue <sup>b</sup>	0.012	Limb/Extremity/Joint Pain <sup>c</sup>	0.074				
8	Anxiety <sup>c</sup>	0.011	Sleep Disorders <sup>c</sup>	0.066				
9	Ischemic Heart Disease <sup>c</sup>	0.009	Abdominal and Bowel Pain <sup>c</sup>	0.065				
10	Lifestyle/life management factors <sup>c</sup>	0.009	Chronic Eye disorders <sup>c</sup>	0.057				

CCSR = Clinical Classifications Software Refined, PCP = Primary Care Provider visits, RUS = Random Undersampling

dCollected in the 6 months prior to each 90-day time interval

# Prediction of Cannabis Use Disorder within the next 90 days: Time-dependent AUC and Brier scores for the top performing model

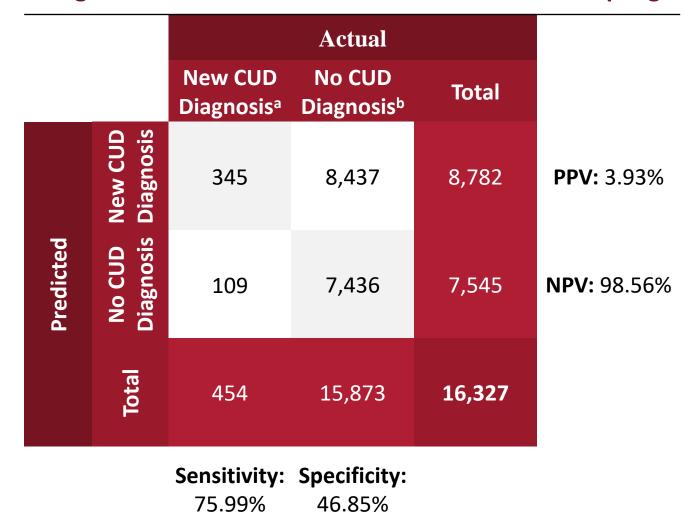
### Logistic Regression-trained model with Random Undersampling 1:10



C/D AUC = Cumulative sensitivity/dynamic specificity area under the receiver-operating characteristic

# Prediction of Cannabis Use Disorder within the next 90 days: Confusion Matrix for the top performing model

**Logistic Regression-trained model with Random Undersampling 1:10** 



CUD = Cannabis Use Disorder, NPV = Negative Predictive Value, PPV = Positive Predictive Value

<sup>a</sup>Subjects with an actual new CUD diagnosis during follow-up were labelled as a true positive if they were predicted positive in the time interval the actual new CUD diagnosis was received. Otherwise, actual positive subjects were labelled as false negative.

bSubjects with no actual new CUD diagnosis during follow-up were labelled as a false positive if they were predicted positive in ≥1 time interval(s) they remained at risk. Otherwise, actual negative subjects were labelled as true negative.

#### **CONCLUSION**

- ✓ A discrete-time-updating prediction tool achieved high levels of discrimination for predicting a new CUD diagnosis among Arkansas MMJ cardholders.
  - ✓ The logistic regression-trained model with 1:10 random undersampling achieved the highest discrimination in predicting CUD in the next 90 days (mean C/D AUC = 0.8068) while the support vector machine-trained model achieved the highest calibration (mean Brier Score = 0.0048)
- This tool could be used to identify persons authorized to purchase MMJ that are at risk of developing CUD to develop targeted screening and treatment programs.

# References

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  4. Arkansas All-Payer Claims Database. Welcome to the Arkansas All-Payer Claims Database (APCD). https://www.arkansasapcd.net/Home/

<sup>&</sup>lt;sup>a</sup>Time interval included for calculation of time-dependent survival probabilities

<sup>&</sup>lt;sup>b</sup>Time interval included as feature

<sup>&</sup>lt;sup>a</sup>Prognostic features are outlined in **BOLD**<sup>b</sup>Acute conditions recorded in the 6 months prior to each 90-day time interval

<sup>&</sup>lt;sup>c</sup>Non-acute conditions recorded anytime prior to the current 90-day time interval