



# Predicting Who Will Seek Mental Health Treatment in the Tech Workplace

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# May is Mental Health Month

## Mental Health Facts IN AMERICA

**Fact:** 43.8 million adults experience mental illness in a given year.



1 in 5 adults in America experience a mental illness.



Nearly 1 in 25 (10 million) adults in America live with a serious mental illness.



One-half of all chronic mental illness begins by the age of 14; three-quarters by the age of 24.

# Context

- ✦ Prevalence
  - ✦ National Alliance on Mental Illness (NAMI) and National Institute of Mental Health (NIMH) reported that 60% of adults with mental illness received no treatment services in the past year
  - ✦ In 2015, estimated 43.4 million (17.9% of U.S. adults) had a mental illness (NIMH, 2015)

# Context

- ✦ Impact
  - ✦ CDC (2017): From 1981 to 2015, suicide is tenth leading cause of death for all adults
  - ✦ Insel (2008): In 2002 alone, Serious Mental Illness cost \$317.6 billion in disability benefits, healthcare expenditures, and lost earnings

# Context

- ✦ Impact in the workplace ([Mental Health America, 2017](#))
  - ✦ Depression as most common and costly disorder
  - ✦ Employees may not seek out treatment due to fear of effect on job



# Dataset

- ✦ Open Sourcing Mental Illness (OSMI) 2014 Mental Health in Tech Survey
  - ✦ Believed to be largest survey on mental health in tech industry at that time
  - ✦ Measures “attitudes toward mental health and frequency of mental health disorders in the tech workplace”

# Research Question

What factors predict whether tech employees will seek mental health services?



# Variables

	Timestamp	Age	Gender	Country	State	Self-Employed	Family History	Work Interfere	Treatment
Data Type	Timestamp	Integer	String	String	String	Binary	Binary	Categorical	Binary
Has NaNs?					Yes	Yes			
Cleaned?		Yes	No*					No*	
<i>N</i>	1259	1259	1259	1259	744	1241	1259	995	1259



# Variables

[illegible][illegible]

# Initial Model

- ✦  $N = 989$ 
  - ✦ Dropped observations with ages below 18 and over 200
  - ✦ Dropped observations containing NaNs (work\_interfere)
- ✦ Features = 22
  - ✦ Dropped features with highly imbalanced classes (Gender, Country, self\_employed)
  - ✦ Dropped features with high frequency of missing data (US State, Comments)
  - ✦ Dropped Timestamp
  - ✦ Added 2 engineered features (US/non-US, Age < 40)
- ✦ Outcome = Have you sought treatment for a mental health condition? (yes/no)

# Initial Model

- ★  $N$  observations = 989,  $p$  predictors = 22

- ★ Applied seven classifier methods:

- ★ Logistic Regression (LR)

- ★ Linear Discriminant Analysis (LDA)

- ★ K-Nearest Neighbor Classifier (KNN)

- ★ Decision Tree Classifier (DTC)

- ★ Random Forest Classifier (RFC)

- ★ Naive Bayes (NB)

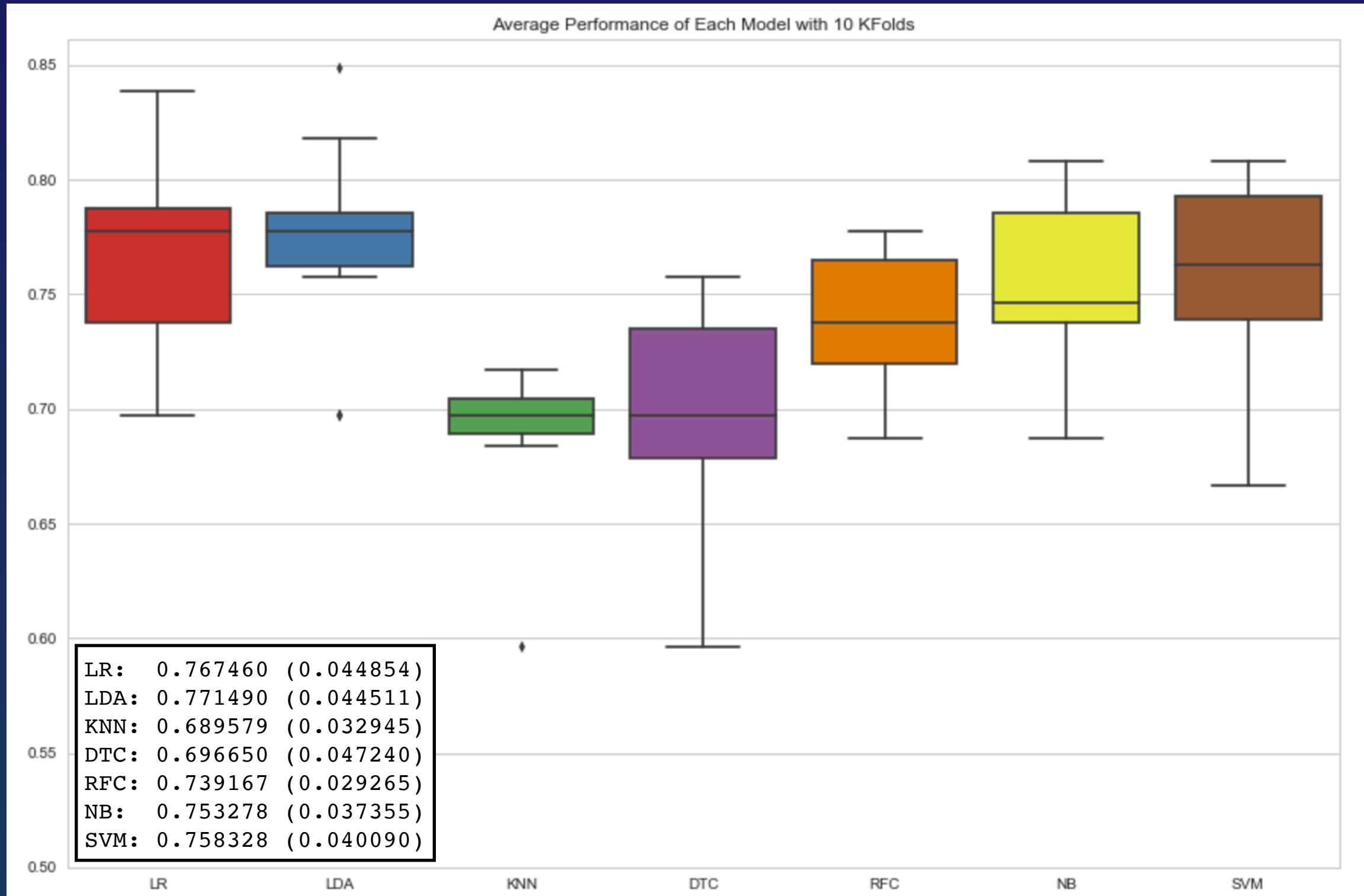
- ★ Support Vector Machines Classifier (SVM)

- ★ Used 10 k-folds cross-validation with mean accuracy (standard deviation accuracy)

LR:	0.767460	(0.044854)
LDA:	0.771490	(0.044511)
KNN:	0.689579	(0.032945)
DTC:	0.696650	(0.047240)
RFC:	0.739167	(0.029265)
NB:	0.753278	(0.037355)
SVM:	0.758328	(0.040090)



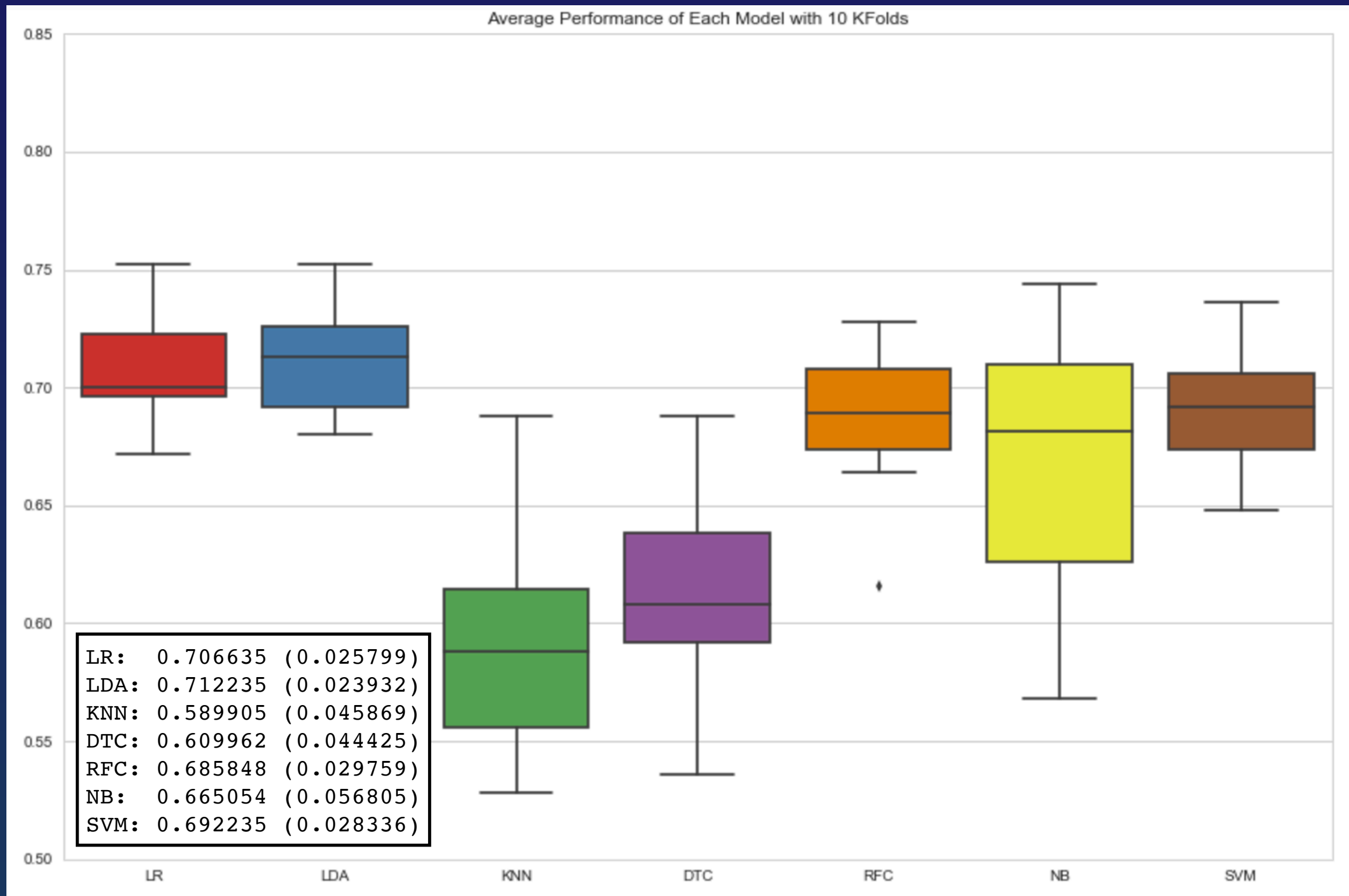
# Initial Model



# Revised Initial Model

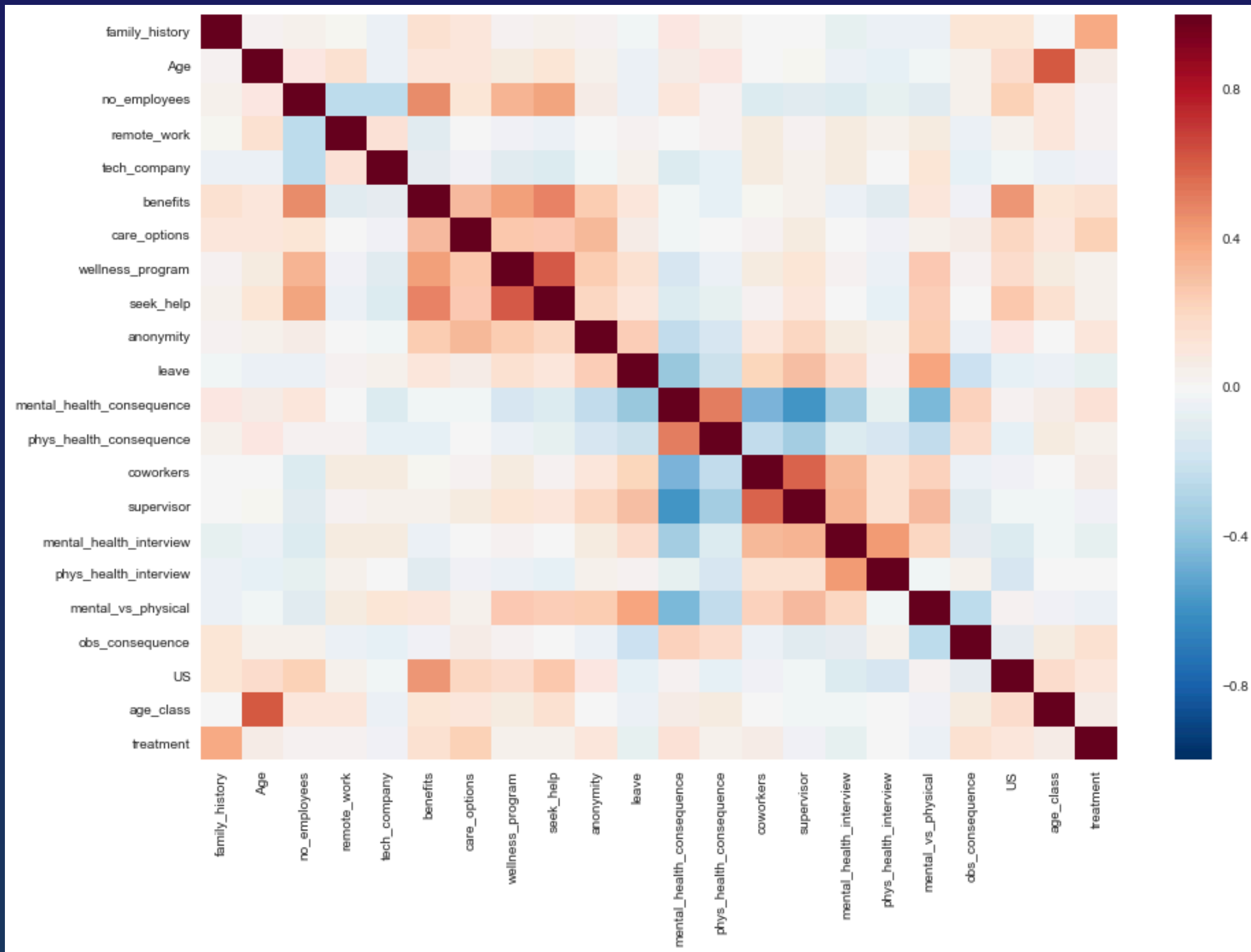
- ✦  $N = 1251$ 
  - ✦ Removed work\_interfere feature which increased  $N$
- ✦ Features = 21
  - ✦ Dropped work\_interfere feature (264 N/As):
    - ✦ If you have a mental health condition, do you feel that it interferes with your work? (Never, Rarely, Sometimes, Often)
    - ✦ Item is confusing and potentially misleading
    - ✦ Another way of stating outcome variable, or confirming participant mental illness?
- ✦ Outcome = Have you sought treatment for a mental health condition? (yes/no)

# Revised Initial Model





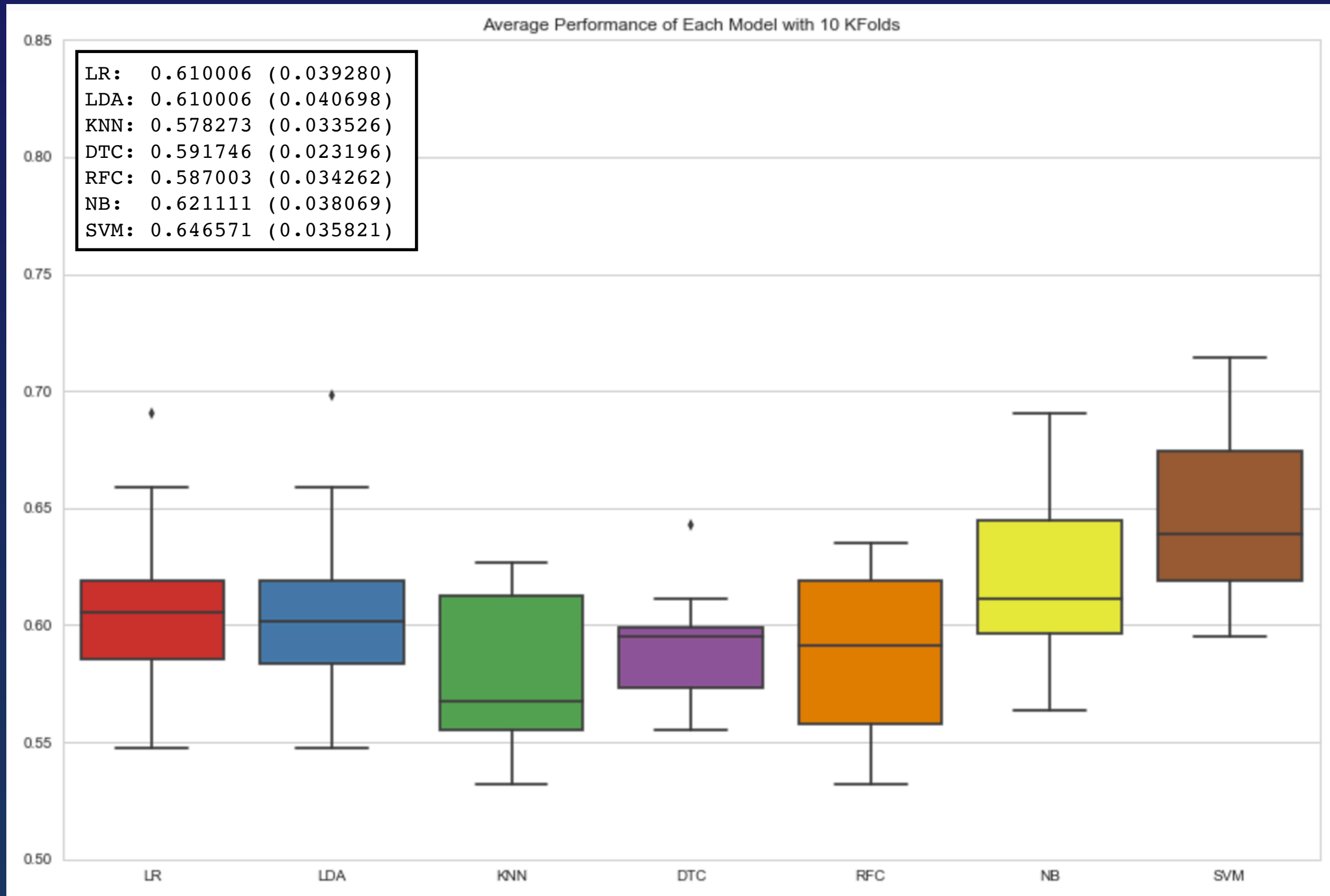
# Feature Reduction



# Reduced-Features Model

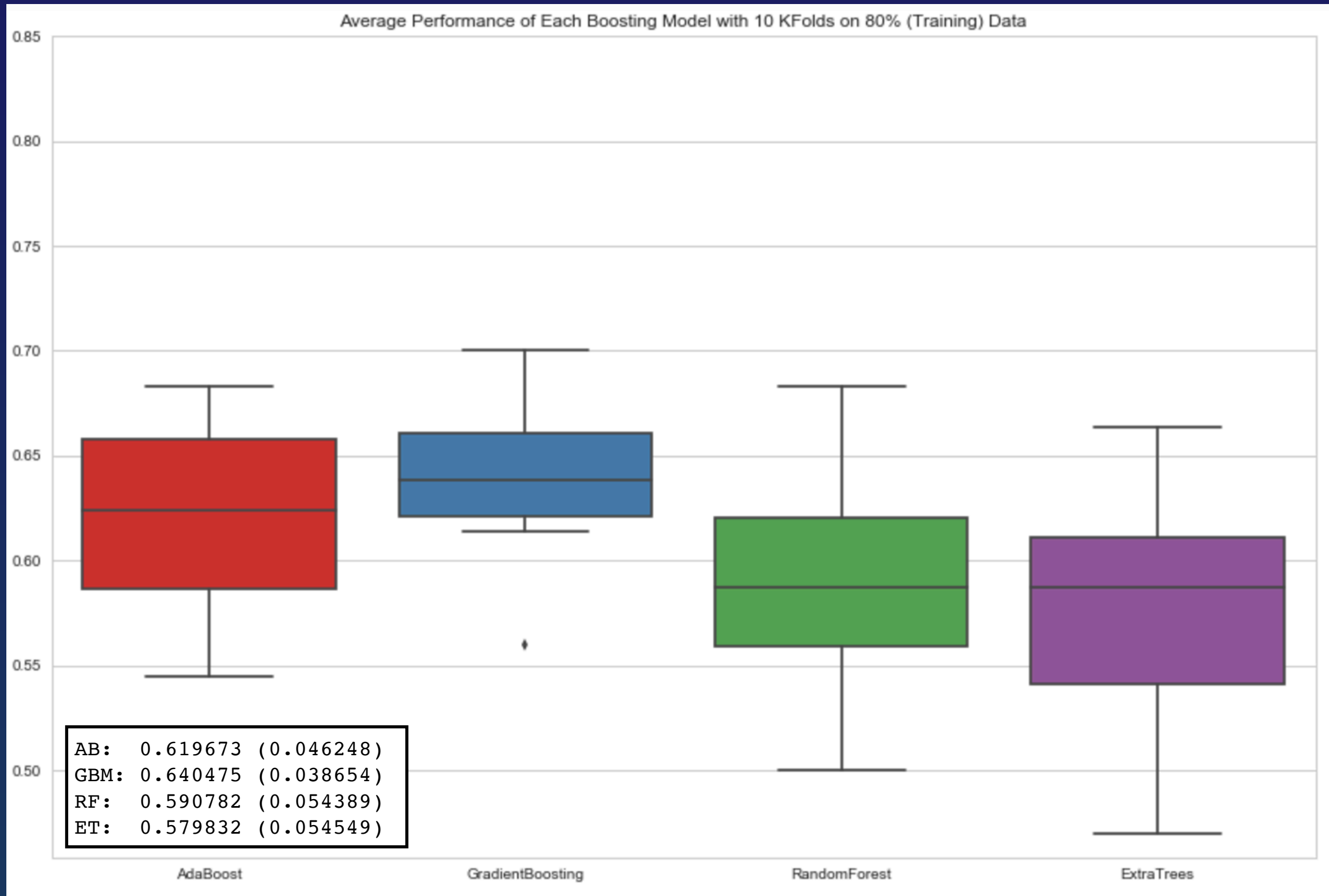
- ★  $N = 1259$ 
  - ★ Maximum  $N$ : Used only features that all participants answered
- ★ Features = 6
  - ★ (1) Number of employees, (2) ease of taking MH leave, (3) knowledge about available MH care options, (4) MH benefits, (5) availability of workplace wellness program, (6) employer-provided resources to seek help
  - ★ Includes generally “actionable” features (e.g., employee education, workplace factors)
- ★ Outcome = Have you sought treatment for a mental health condition?

# Reduced-Features Model





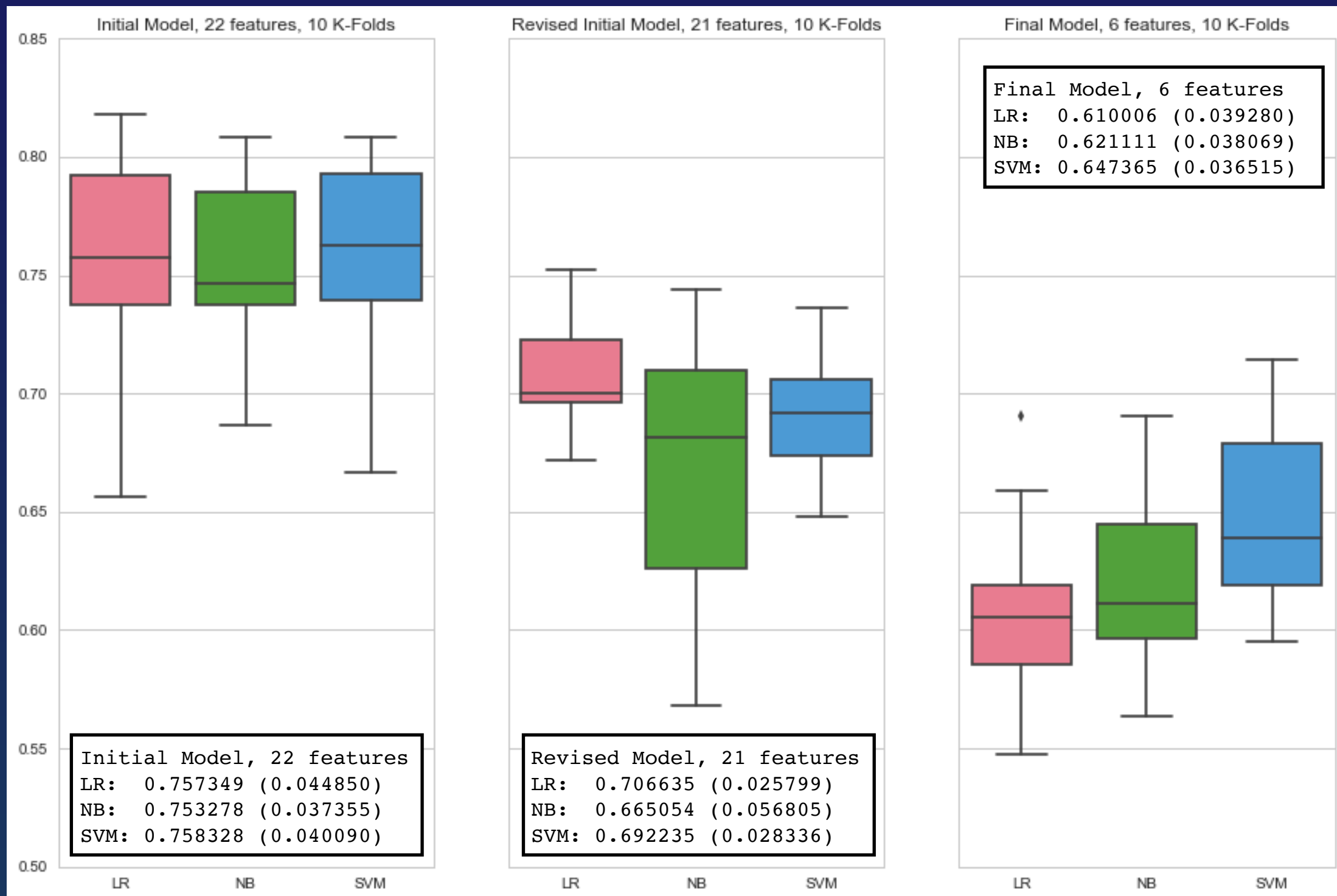
# Boosting Models



# Hyperparameter Tuning

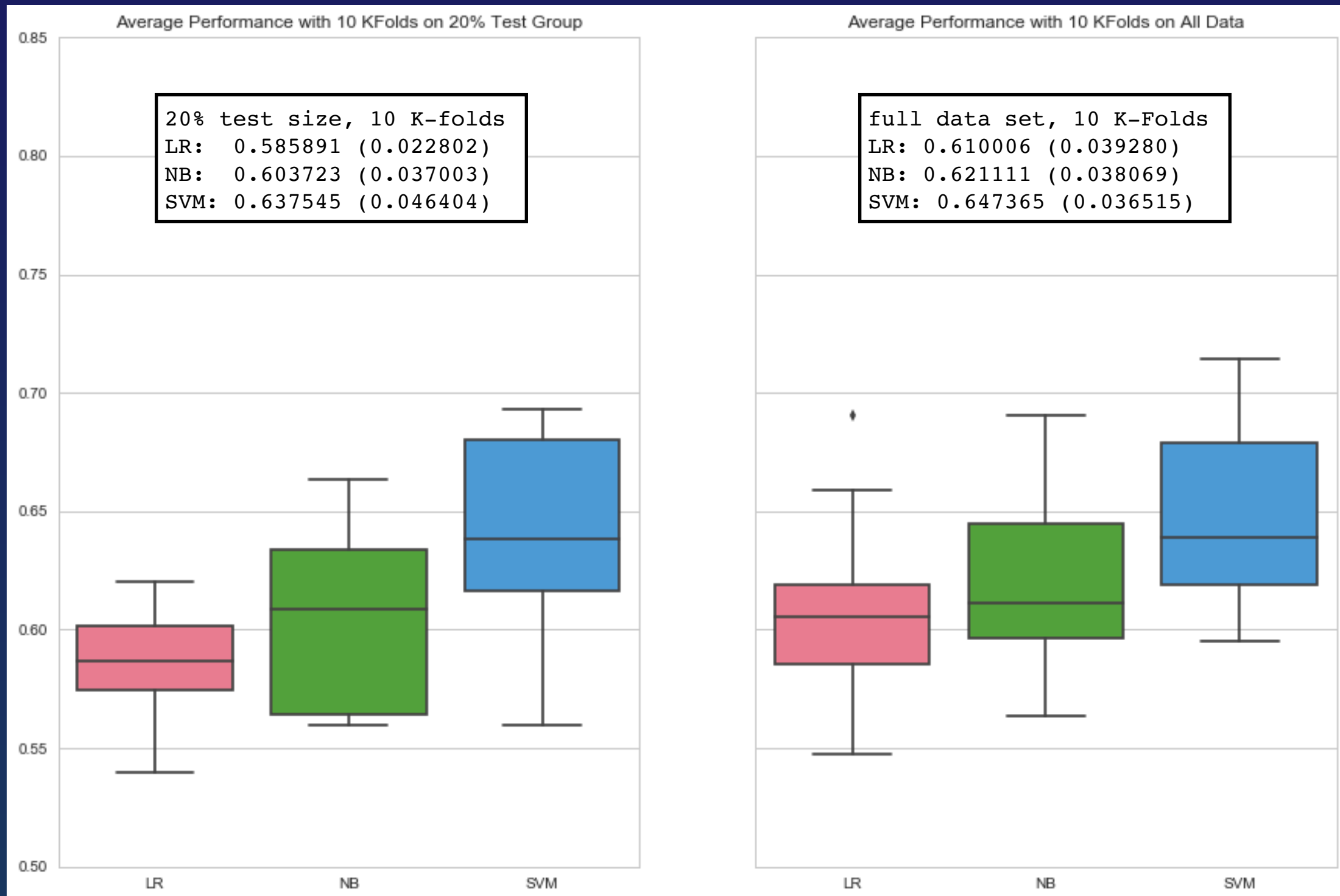
- ✦ Tuned select classifiers in reduced-features model:
  - ✦ Logistic Regression
  - ✦ Random Forest
  - ✦ SVM Classifier
- ✦ Minimal increases in performance
- ✦ Kept Logit, Naive Bayes, and SVM Classifier for final model

# Comparison of Initial, Revised, and Reduced-Features Models





# Test Group vs. Full Dataset



# Outcome

The size of the organization, ease of taking leave, employees' knowledge about available mental health care options, availability of mental health benefits, availability of a workplace wellness program, and employee-provided mental health resources may predict which tech employees will seek mental health services with approximately 60-65% accuracy.

# Methodological Concerns

- ✦ Directionality/Causality
  - ✦ Does seeking mental health services predict having workplace benefits? Vice versa?
- ✦ Survey Items
  - ✦ Comments from participants about needing N/A or I don't Know for certain items (e.g., family history)
  - ✦ Words used interchangeably: Condition, Diagnosis, Illness

# Extensions

- ✦ Compare to OSMI's 2016 dataset (data collection still in progress!)
  - ✦ Validate model against similar outcome variable
  - ✦ Correlate responses using Likert-type scales
- ✦ Perform Text Analysis
  - ✦ Employ other text analyses (e.g., n-grams, NLP)
  - ✦ Determine how text supports or fails to support model

# Conclusion

- ✦ Providing employee education about mental health care options and offering a workplace wellness program may predict tech employees' behaviors to seek mental health treatment
- ✦ Other factors such as the size of the organization and availability of mental health benefits may also predict which tech employees will seek mental health treatment
- ✦ These findings may or may not be representative for employees in other fields



# Questions?

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Treatment in the Tech Workplace

[https://github.com/KevinHahnMTBC/  
Mental-Health-and-Tech](https://github.com/KevinHahnMTBC/Mental-Health-and-Tech)