# SMOKING REDUCTION TRAJECTORIES AND THEIR ASSOCIATION WITH SMOKING CESSATION

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## Introduction

- Smoking is the leading cause of premature death and preventable illness worldwide [7].
- Reduction may be as effective abrupt quitting in achieving cessation [4].
- When people are asked to reduce smoking, how do people choose to do so?
- Are there smoking or demographics associated with certain reduction patterns?
- Which patterns of reduction are associated with better cessation outcomes?

#### Methods

#### Data

- $\bullet$  Smoking and demographic information from 5 clinical trials of NRT [1–3, 5, 6]
- Baseline and follow-up (weeks 2, 10, 18, and 26) CPD were recorded
- CPD and expired breath carbon monoxide (CO) collected at week 56
- In the parent trial:
- -Participants were randomly assigned to receive active or placebo NRT
- -All participants were told to reduce their smoking as much as possible

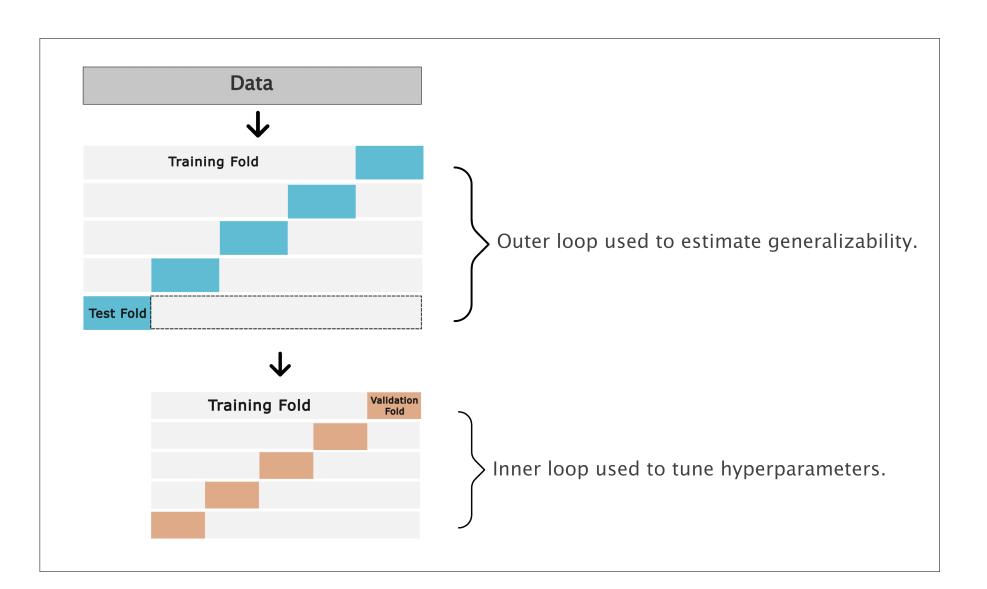


Fig. 1: Nested cross validation scheme. Data were partitioned into 80% training and 20% testing sets to evaluate generalizability across five folds. Each training fold was further divided into 80% training and 20% validation folds to tune hyperparameters.

#### Analysis

- 1. We estimated smoking trajectories using latent class analysis (LCA) as a function of percent reduction in CPD. Participants were assigned to the most likely latent class.
- 2. We used regularized regression (i.e., elastic net) under a nested cross validation scheme (see Figure 1) to predict latent class using baseline and demographic characteristics.
- 3. We predicted biochemically-verified smoking status at week 52 using baseline and demographic characteristics, plus latent class.
- Pre-registered protocol: https://osf.io/qh378/
- Analytical code: https://github.com/ajbarrows/mcneil-lca

#### Results

## **Participants**

- 108/2066 participants were excluded for missing values.
- Resulting n = 1783:
- From five countries
- -44.8% male, mean age  $44.10 \pm 10.72$  years
- -Smoked an average of  $27.32 \pm 9.73$  CPD

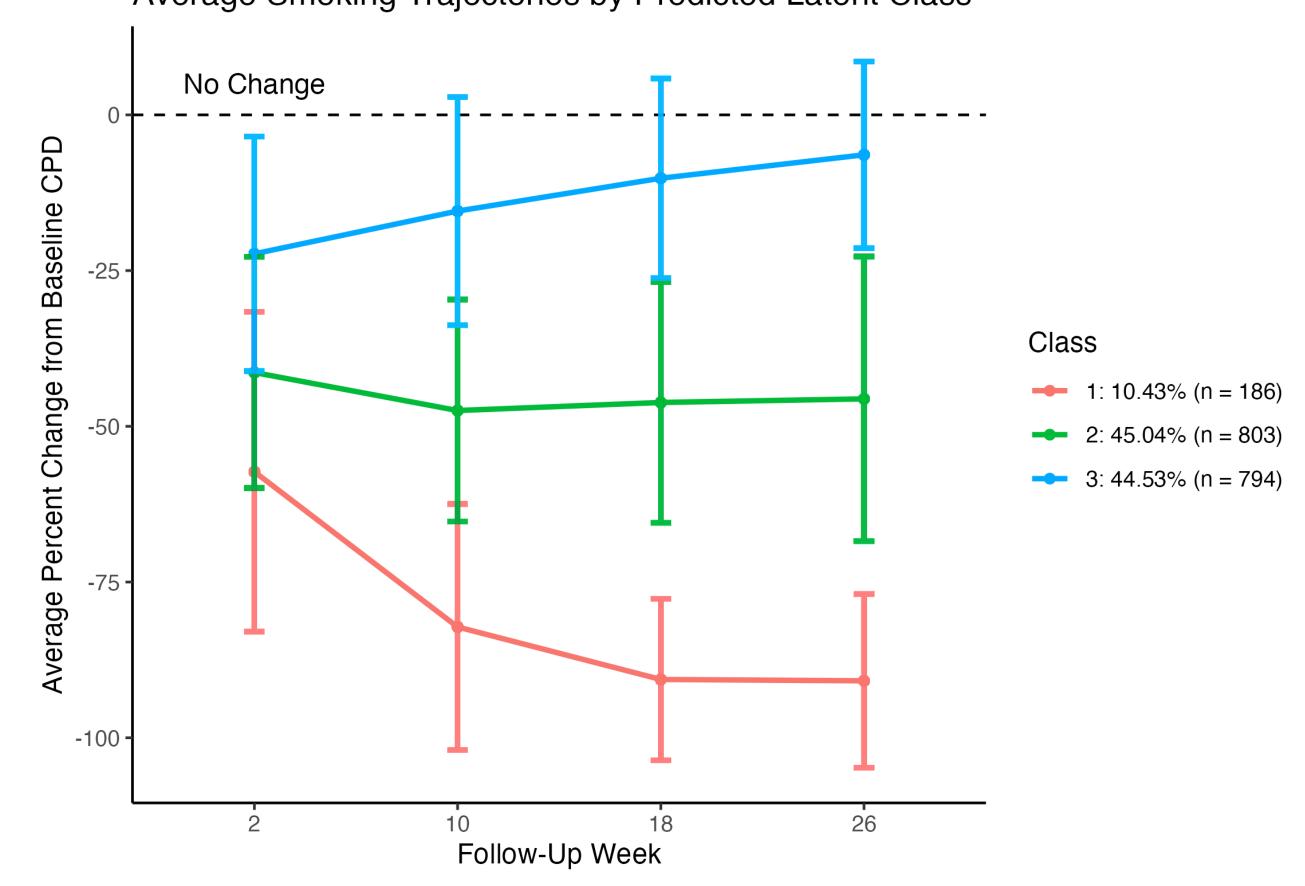
## Latent Class Analysis

- Class 1:  $\sim 10\%$  initially reduced and nearly eliminated smoking
- Class 2:  $\sim 45\%$  reduced by nearly 50% and remained
- Class 3:  $\sim 45\%$  initially reduced but reverted to their baseline smoking

## **Predicting Latent Class**

- Demographic data and baseline characteristics (e.g., smoking and quit behavior, FTND, SF-36, trial treatment group) were used as independent variables
- Latent class was used as the dependent variable
- One cross-validated elastic net logistic regression for each latent class (i.e., one-versus-all)
- All models performed better than chance:
- -Class 1 test AUC = 0.766, p < .001. Tended to be older with lower anxiety and nicotine dependence, more likely to have received active NRT.
- -Class 2 test AUC = 0.569, p = .008 No clear pattern of characteristics.
- -Class 3 test AUC = 0.523, p < .001 Inverse of class one: higher nicotine dependence, more likely to have received placebo NRT.

## Average Smoking Trajectories by Predicted Latent Class



## Predicting Smoking Status

- 122/1783 (6.8%) achieved biochemically-verifiable smoking cessation at week 52:
- -Class 1: 70/186 (37.6%); Class 2: 34/803 (4.2%); Class 3: 18/776 (2.3%)
- Elastic net logistic regression predicting smoking cessation using
- baseline characteristics alone: AUC =  $0.632 \pm 0.006$ , p < .001– baseline characteristics plus latent class: AUC =  $0.776 \pm 0.010$ , p < .001
- Adding latent class as an independent variable improved cessation prediction by 14.4%

## Conclusions

- Examining latent trajectories in smoking behavior among people not motivated to quit revealed three distinct patterns
- One of these trajectories was nearly twice as likely as the others to achieve cessation
- Smoking reduction in the first two weeks after intervention by  $\geq 50\% \rightarrow$  substantially increased cessation likelihood

# Acknowledgements

## This work is supported by

- NIH (NIDA T32DA045593 and NIGMS P20GM103644)
- Cancer Research UK (PRCPJT-Nov22/100012, PPRCPJT\100,023, and C56067/A21330)
- The University of Oxford and NHS Greater Manchester Integrated Care

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