

Impact Of Variable Selection on Mode Effect Adjustment for A Longitudinal Study

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- > Surveys using multiple modes of data collection are increasingly popular
- > Concerns about potential mode effects in multimode surveys
- > Mode of data collection affects resultant survey estimates through
 - Selection effect due to different respondents choosing different mode
 - E.g., in a sequential web-mail survey, younger people are more likely to choose the web mode whereas older people are more likely to choose the paper mode
 - Measurement effect due to respondents providing different answers to different mode
 - E.g., respondents are more likely to provide socially desirable answers to an intervieweradministered mode than a self-administered mode (Tourangeau and Yan, 2007)
 - An unknown mix

> It is critical to understand, estimate, and adjust for mode effects in a multimode survey

- > The regression modeling approach is one approach to estimate and adjust for mode effects (Kolenikov and Kennedy, 2014)
 - Regressing survey responses on mode, demographic variables, and other related variables
- > This talk evaluates the impact of variable selection in the regression modeling approach on mode effect adjustment

> Synthesized a longitudinal dataset including

- Mode: By Phone vs. ACASI
- Binary outcomes (past month tobacco use) measured at the current wave (time t)
 - Mode effects in these outcomes to be evaluated and adjusted
- Variables to be included in the regression model
 - Model 1: Mode + demographic variables
 - Model 2: Model 1 + outcome measured at one prior wave of data collection (time t-1)
 - Model 3: Model 2 + one binary indicator of whether outcome changed between time t-2 and time t-1
 - Model 4: Model 2 + a count variable indicating the number of times outcome changed between time t-4 and t-1
- Models run for adults (26+), and young adults (18-25)

> Over 14,000 completes, Phone vs. ACASI (75% vs. 25%)

> Demographic characteristics

			Outco	omes	
	Mode	Cigarette	Vaping	Cigar	Smokeless
Age Category	***	***	***	*	***
Sex	*			***	***
Race/Ethnicity		***	***	***	***
Marital Status	**	***	***	***	***
Education Level	***	***	***	***	***
Income Level	***	***	***	***	
Working Status	**	***	***	***	
Region	***	***	*		

Note: * p < .05; ** p < .01; *** p < .001

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> Outcome related variables

	Mode				Outcomes				
	Cigarette	Vaping	Cigar	Smoke less	Cigarette	Vaping	Cigar	Smoke less	
Outcome at t-1	***		*		***	***	***	***	
Change t-1 vs t-2		**	*		**	***	***	***	
Change t-1 to t-4		*	*		**	***	***	***	

> With demographic characteristics

	Cigarette	Vaping	Cigar	Smokeless
Age Category	**	*		
Sex	*		*	***
Marital Status	*			
Education Level			*	

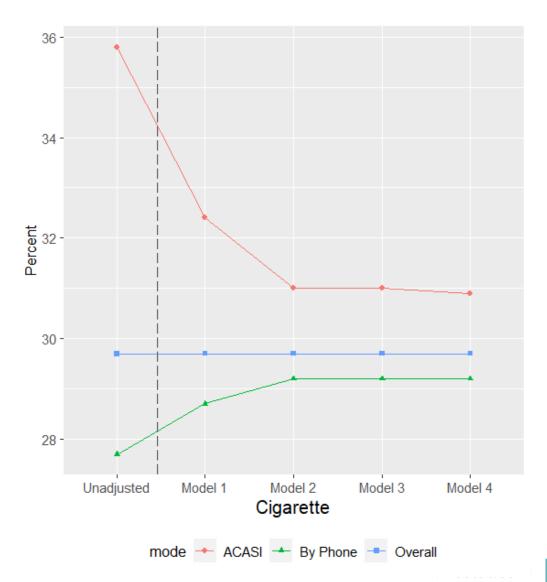
- > With outcome related variables
 - No significant interactions

Results: Adults – Model Comparison and Estimate Adjustment

	-			
	Cigarette			
Unadjusted Mode Difference	***			
Model Comparison	Pseudo R ²	LRT		
Model 1 (Mode + Demog.)	0.149			
Model 2 (M1 + time t-1)	0.500	***		
Model 3 (M2 + t-1 vs. t-2)	0.501	***		
Model 4 (M2 + t-1 to t-4)	0.500	***		

Note: * p < .05; ** p < .01; *** p < .001

> Past Month Use: Cigarette

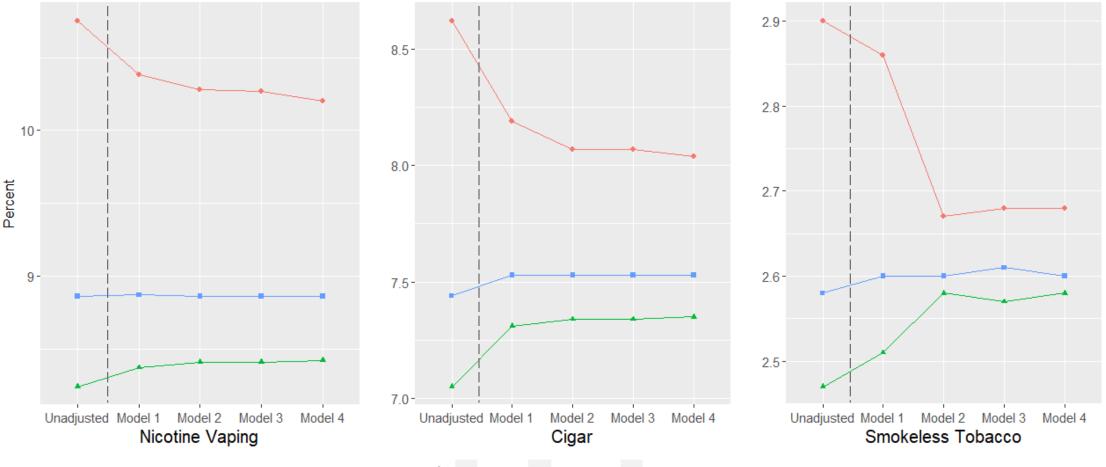


> Past Month Use: Other tobacco products

	Vaping		Ciga	r	Smokeless		
Unadjusted Mode Difference	***		**				
Model Comparison	Pseudo R ² LRT		Pseudo R ² LRT		Pseudo R ² LRT		
Model 1 (Mode + Demog.)	0.043		0.025		0.040		
Model 2 (M1 + time t-1)	0.133	***	0.116	***	0.132	***	
Model 3 (M2 + t-1 vs. t-2)	0.134	**	0.116		0.133	**	
Model 4 (M2 + t-1 to t-4)	0.137	***	0.117	***	0.132		

Results: Adults – Model Comparison and Estimate Adjustment (3)

> Estimates of Past Month Use: Other Tobacco Products

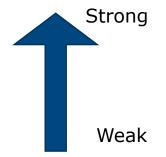


mode 🔶 ACASI 📥 By Phone 💶 Overall

Adults: Summary

> Pattern of mode effect and its adjustment

- Use rate as reported in ACASI is always higher than that reported by phone
- The gap between the outcome estimates by mode narrowed after adjustment
- Overall estimate somehow was not affected by the variables added into the model
- > Impact of variables selected to the model
 - Outcome measured at one prior wave
 - Demographic variables measured at the same wave
 - Number of changes through multiple time points
 - Change between the last two waves



> Over 9,300 completes, Phone vs. ACASI (80% vs. 20%)

> Demographic characteristics

			Outc	omes	
	Mode	Cigarette	Vaping	Cigar	Smokeless
Age Category	*	***	*	*	
Sex		*			**
Race/Ethnicity	***	**	***		
Marital Status	**		*		
Education Level	***	***		***	
Income Level	***	***			
Working Status		***	***	**	
Region	***	***	***	**	

> Outcome related variables

	Mode				Outcomes			
	Cigarette	Vaping	Cigar	Smoke less	Cigarette	Vaping	Cigar	Smoke less
Outcome at t-1	**				***	***	***	***
Change t-1 vs t-2	*				**	***	***	***
Change t-1 to t-4					***	***	***	***

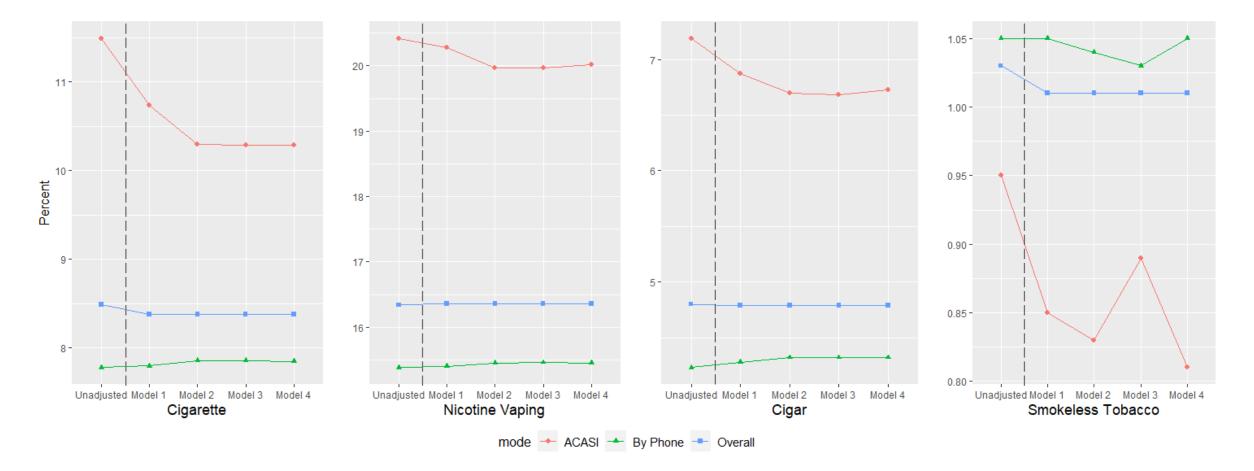
> Interactions with Mode

		Past Month Use						
	Cigarette	Vaping	Cigar	Smokeless				
Demographic characteristics								
Age Category		**		*				
Race/Ethnicity	**							
Education Level	***							
Region		**						
Outcome related variable	es							
Outcome at t-1				*				

> Past Month Use

	Cigarette		Vaping		Cigar		Smokeless	
Unadjusted Mode Difference	***		***		***			
Model Comparison	Pseudo R ²	LRT						
Model 1 (Mode + Demog.)	0.035		0.024		0.009		0.005	
Model 2 (M1 + time t-1)	0.129	***	0.152	***	0.049	***	0.041	***
Model 3 (M2 + t-1 vs. t-2)	0.129		0.152		0.050		0.042	
Model 4 (M2 + t-1 to t-4)	0.130	**	0.155	***	0.050	**	0.041	

> Estimates of Tobacco Product Use in Past Month



Young Adults: Summary

> Pattern of mode effect adjustment

- The gap between the outcome estimates by mode narrowed after adjustment but the magnitude is smaller than adults
- > Impact of variables selected to the model
 - The effect of outcome measured at one prior wave is not significantly larger than the joint effect of the demographic variables measured at the same wave
 - Change between the last two waves didn't contribute to the mode adjustment among young adults

Conclusion

> Findings

- Using the longitudinal information helped to narrow the gap between mode, especially the most recent information
 - Even if no mode difference on outcomes
 - Age difference on the mode effect adjustment
- No apparent adjustment on the overall estimates
 - ACASI was affected more by adjustment, but the phone completes dominated the distribution
 - Missing variable(s) that are more informative in mode adjustment

Conclusion (2)

> Limitation

- Lake of randomization in the mode assignment
 - Selection effect
 - Measurement effect
- Missing sample weight
 - hard to compare the adjusted values with the results from external surveys

> Future research

- Search for more informative variables for mode effect adjustment
- Other mode adjustment methods
 - Propensity matching / weighting
 - Imputation

- > Kolenikov S., and Kennedy C. (2014). Evaluating Three Approaches to Statistically Adjust for Mode Effects, *Journal of Survey Statistics and Methodology*, 2(2): 126–158. <u>https://doi.org/10.1093/jssam/smu004</u>
- Tourangeau, R., and Yan, T. (2007). Sensitive Questions in Surveys. *Psychological Bulletin*, 133(5):859-83. doi: 10.1037/0033-2909.133.5.859.





Thank you!

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