Using Latent Class Probability Estimation and Residual Inclusion to Address Confounding in Medication Adherence Modeling

Julia F. Slejko, PhD¹ Louis P. Garrison, PhD¹ Richard J. Willke, PhD² ¹University of Washington Pharmaceutical Outcomes Research and Policy Program ²Pfizer, Inc.

Background: Medication Adherence

- Often suboptimal and variable in chronic disease settings.
- Using claims data, we find some patient characteristics are associated with adherence.
- Patterns of adherence may also be associated with characteristics that we cannot observe.



•Difficult to estimate the association between adherence and outcomes, partly due to potential confounding by unobserved patient characteristics.



Objectives

• To characterize variation and predictors of medication adherence.

 To improve the ability to explain the effect of adherence on outcomes by adjusting for unobserved confounders.



Methods

- Cohort definition, adherence estimation
- Three-part analysis:
 - 1. Characterize adherence variation
 - 2. Estimate probability of adherence
 - 3. Associate adherence and outcomes

"Real-World" Adherence

- 10% random sample of the IMS LifeLink Integrated Patient-Centric Claims (5.6 million), 1997-2008
- Adult new statin users for primary prevention
 - In prior 12 months, no cardiovascular diagnoses or statin use (n=20,858)
- Minimum two-year follow-up
- Adherence measured as yearly <u>Proportion of</u> <u>Days</u> <u>Covered</u> (PDC)

Adherence and CV risk





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Cohort Characteristics

Characteristic	Study Cohort
Total	N = 20,858
Male n (%)	10,382 (49.77%)
Mean age in years (SD)	55.18 (10.6) (Range: 18-96)
No healthy behaviors n (%)	11,853 (56.83%)
(e.g., flu shot, physical, etc.)	
Chronic disease indicator score (SD)	4.79 (2.92) (Range: 1-23)
(measure of comorbidity using Rx claims)	
Generic statin (vs. brand) n (%)	2,215 (10.62%)
CV Event (MI, stroke) n (%)	1,522 7.30%

Distribution of Adherence: Years 1 & 2



- Appears that high- and low-adherers exist.
- Variation in adherence behavior.

Finite Mixture Model

- Model scenario where >1 distributions exist.
- Allows one to identify and estimate the parameters of interest for each sub-population in the data, not just of the overall mixed population
- Provides a natural representation of unobserved heterogeneity in a finite number of latent classes

Year 2 adherence = $\alpha + \beta_1 male + \beta_2 age + \beta_3 generic statin + \beta_4 healthy behavior + \beta_5 chronic disease indicator + <math>\epsilon$



Part 1: FMM Results

	Change in year 2			
N=20,858	adherence	P>z	[95%Conf.	Interval]
Class 1				
Male	0.032	< 0.001	0.020	0.044
Age	0.003	< 0.001	0.003	0.004
No healthy behaviors	-0.065	<0.001	-0.077	-0.053
CDI	0.000	0.898	-0.002	0.002
Generic	-0.017	0.077	-0.036	0.002
Constant	0.201	0.000	0.169	0.233
Probability	67.41%	0.000	66.39%	68.41%
Class 2				
Male	0.001	0.654	-0.002	0.003
Age	0.001	< 0.001	0.001	0.001
No healthy behaviors	-0.008	< 0.001	-0.010	-0.005
CDI	0.001	0.006	0.000	0.001
Generic	-0.001	0.670	-0.005	0.003
Constant	0.911	< 0.001	0.901	0.920
Probability	32.59%		31.59%	33.61%

Who is in each component?

Classification of subjects based on most likely latent class membership.

N=20,858	Frequency	Percent	Mean Adherence
Class 1 "Low"	13,632	65.37	32.73%
Class 2 "High"	7,224	34.63	95.46%



Determinants of Posterior Probability

Ordinary least squares regression

Class 2 posterior probability = $\alpha + \beta_1 male + \beta_2 age + \beta_3 generic statin + \beta_4 healthy behavior + \beta_5 chronic disease indicator + <math>\epsilon$



Part 2: Results

Determinants of Class 2 posterior probability.

N=20,858	Coef.	P>t	[95% Conf.	Interval]
Male	0.045	<0.001	0.033	0.057
Age	0.005	<0.001	0.005	0.006
No healthy				
behaviors	-0.051	<0.001	-0.062	-0.039
CDI	0.000	0.951	-0.002	0.002
Generic	0.003	0.716	-0.015	0.022
_cons	0.045	0.007	0.012	0.079



Modeling Adherence and CV risk

- Cox proportional hazards model
- Up to 5-year risk of vascular events (MI, Stroke)
 - Follow-up: mean 42 months (range 24-119)
- Excluded patients with Year 1 CV events.

$$\begin{split} &\log h_i CVEvent(t) = \beta_1 Year \ 2 \ Adherence + \beta_2 male + \beta_3 age \\ &+ \beta_4 CDI + \varepsilon \\ &\log h_i CVEvent(t) = \beta_1 Year \ 2 \ Adherence + \beta_2 male + \beta_3 age \\ &+ \beta_4 CDI + \beta_4 Part \ 2 \ residual + \varepsilon \end{split}$$

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Part 3: Cox Proportional Hazards Results

N=20,858	Hazard Ratio	P>z	[95%Conf.	Interval]
Model 1				
Adherence	0.676	<0.001	0.594	0.769
Male	1.238	<0.001	1.117	1.373
Age	1.055	<0.001	1.051	1.060
CDI	1.190	<0.001	1.174	1.206
Model 2				
Adherence	0.530	<0.001	0.427	0.658
Male	1.251	<0.001	1.128	1.387
Age	1.057	<0.001	1.052	1.062
CDI	1.190	<0.001	1.174	1.207
Residual (covariate)	1.328	0.006	1.086	1.624



Limitations

- Using pharmacy claims for adherence estimation.
- Patient censoring.
- Structural uncertainty.



Conclusions

- FMM allows analysis of heterogeneity in statin adherence.
- Patient characteristics are associated with likelihood of adherence.
- Unobserved determinants can be captured and used to investigate true effect of adherence on outcomes.



Future Work

- Model structure.
- Validation data set to investigate residuals.



Questions

slejko@uw.edu