

Realistic policies on influenza vaccination and risk of subsequent influenza among university students: application of targeted maximum likelihood with interference

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¹ Footnotes are reserved for asides or references

Defining Vaccine Effectiveness (VE)

Potential outcomes

$$Y_i(\boldsymbol{v}) = Y_i(v_i, v_{-i})$$

where $\boldsymbol{v} = (v_1, v_2, \dots, v_i, \dots, v_n)$

Under no interference

$$Y_i(v_i, v_{-i}) = Y_i(v_i)$$

Vaccine effectiveness (VE)

- Direct or unit-specific VE
 - Common in both randomized and observational studies²

$$\frac{1}{n} \sum_i^n Y_i(1, v_{-i}) - Y_i(0, v_{-i})$$

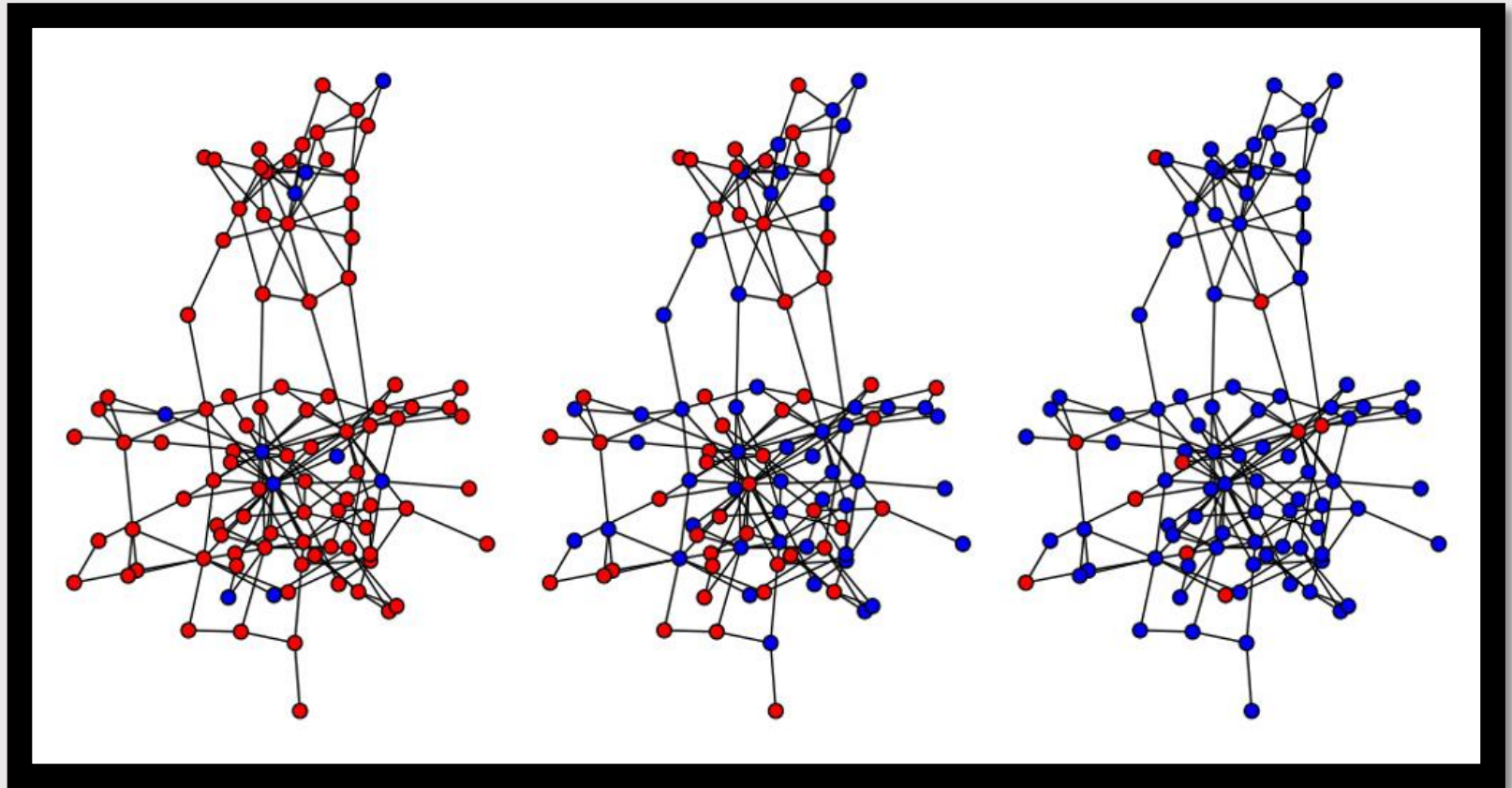
- “What is the expected risk of changing a single person’s vaccination status, holding the vaccination status of the remainder of the population as constant?”

² Patterns of vaccination in observational data can lead to a form of confounding, see Zivich et al. (2021) *Am J Epidemiol* for details and simulations

Alternative definition of VE

- Instead, estimate the risk under varying vaccination distributions

$$\frac{1}{n} \sum_i^n Y_i(\mathbf{v}')$$



Application

Motivating question

Importance of university students

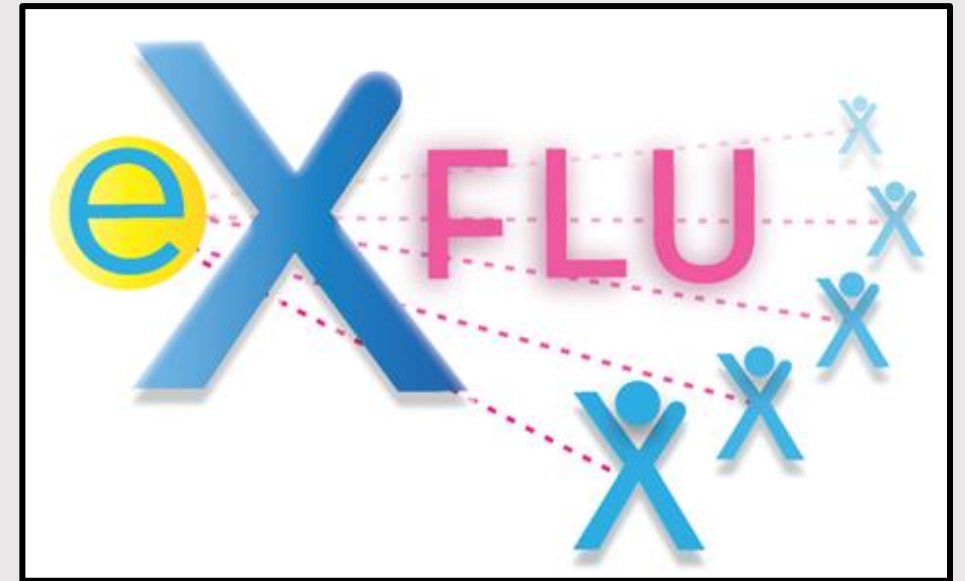
1. Elevated risk of influenza
2. Low vaccination rates
3. Negative impacts on well-being
4. Importance for subsequent transmission

What is the 10-week risk of influenza under different distributions of influenza vaccination (policies)?

- Policies shift the log-odds of vaccination

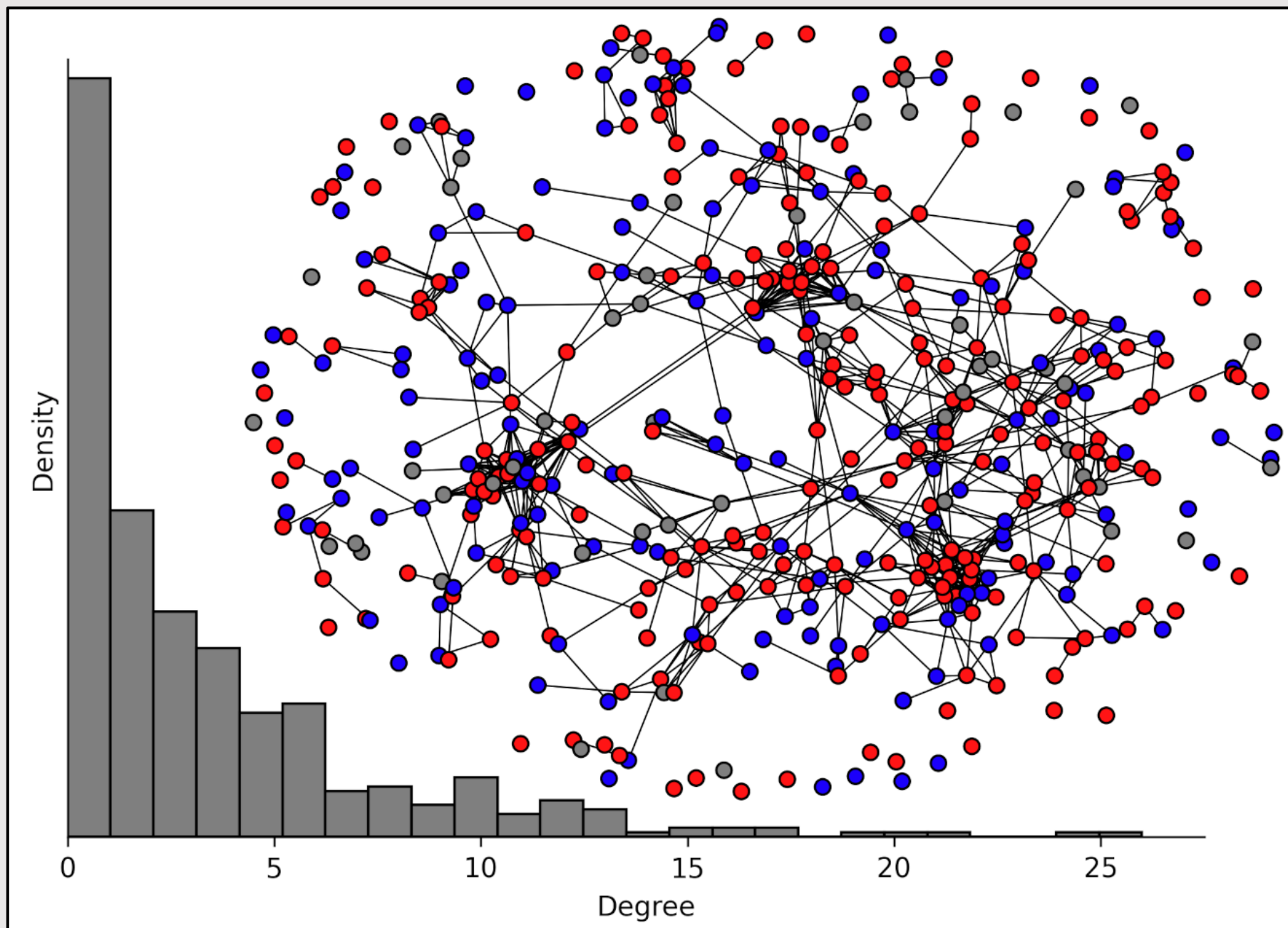
eX-FLU³

- Cluster randomized trial
 - Effect of 3-day self-isolation to mitigate transmission
 - Jan – Apr 2013
 - Midwestern university



Covariates

- Exposure
 - Self-reported vaccination
- Outcome
 - Influenza-like illness (coughing + fever, body aches, or chills)
 - Laboratory-confirmed influenza (PCR)
- Confounders
 - Gender, race, stress, hand hygiene, high-risk conditions, sleep quality, alcohol use, assigned intervention



Measurement error

Self-reported weekly contacts prone to measurement errors

Bayesian network measurement error correction⁴

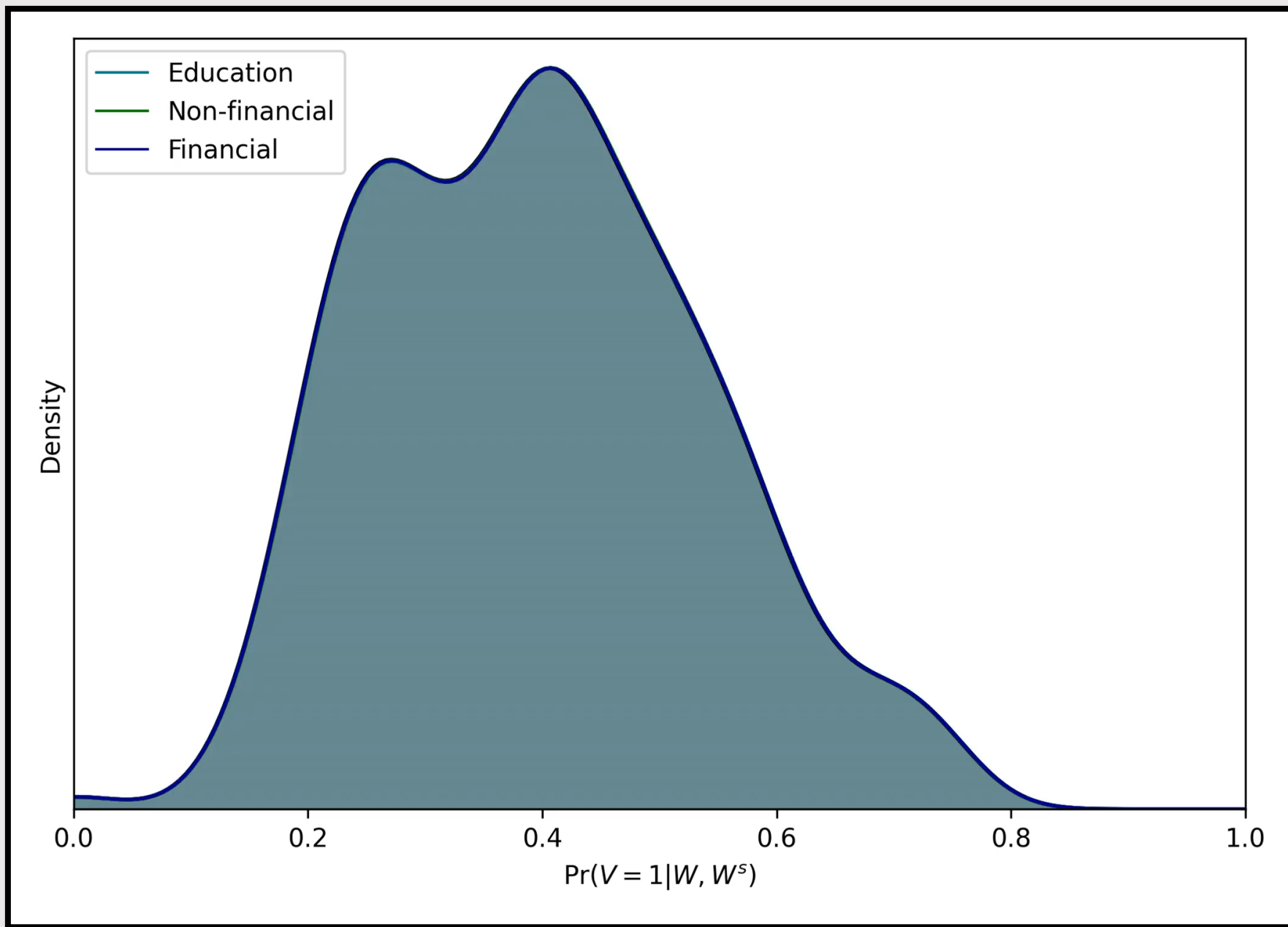
- Sensitivity and specificity
 - Subsample (n=103) had Bluetooth-collected contacts
- Estimate probability of edges between all pairs
- Multiple imputation for measurement error correction
 - Generated 100 networks

⁴ Young et al. (2020) *Journal of Complex Networks*

Policies

- Three categories of policies
 - Based on self-reported reasons for unvaccinated
 1. **Educational** (dispel myths and highlight benefits)
 2. **Non-financial barriers** (ease access and availability)
 3. **Financial barriers** (reduce cost)
- Shift in log-odds of vaccination
 - Shift by ω if targeted, $\omega/3$ otherwise
 - $\omega = (0, 0.25, 0.50, \dots, 3)$

	Unvaccinated (n=241, 53%)
Educational	50 (21%)
Non-financial	93 (39%)
Financial	18 (8%)
Allergy	2 (1%)



Estimation

Targeted maximum likelihood estimation for network data

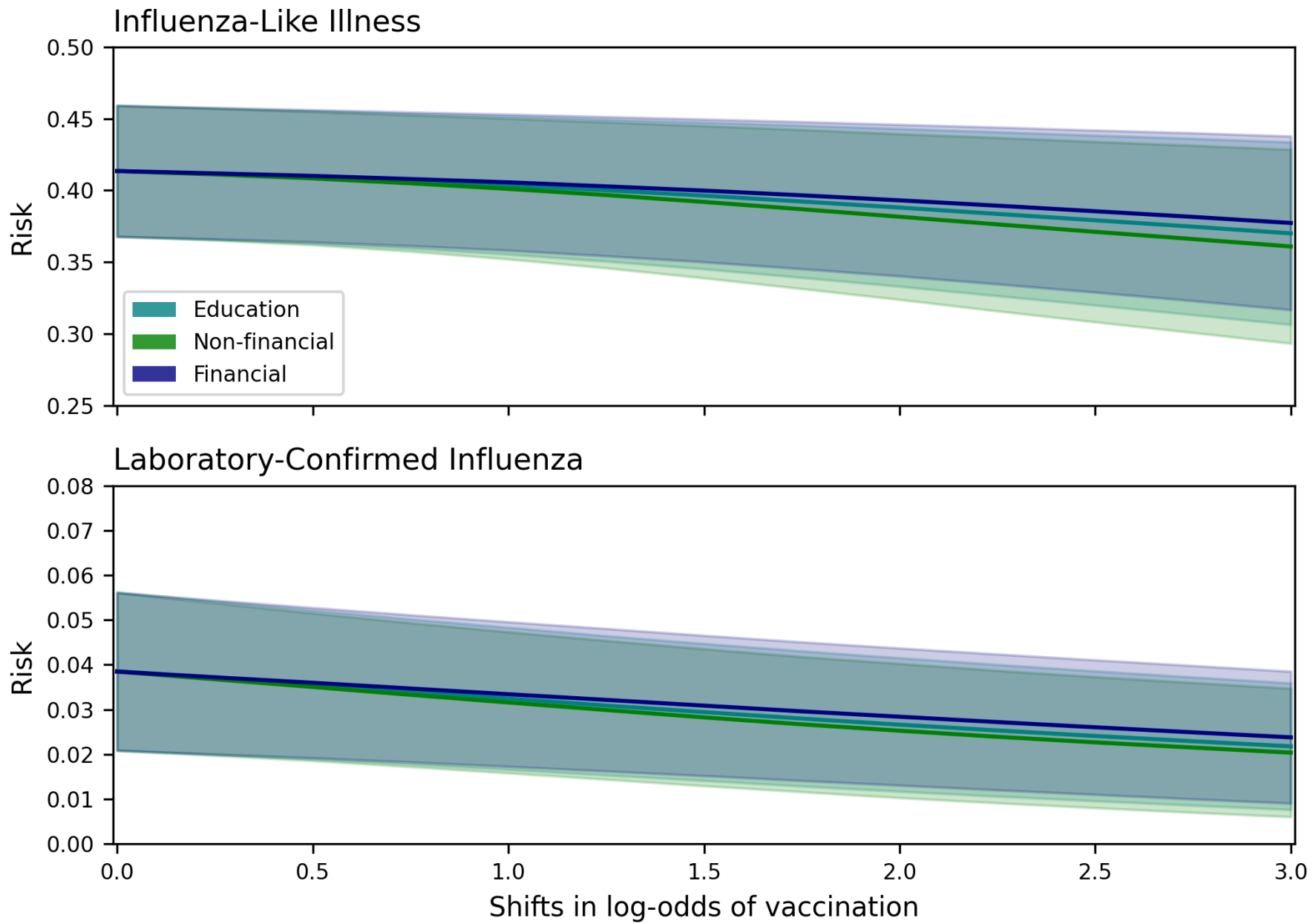
- Allows for dependence due to immediate contacts in network
 - Dependence due to vaccination and other covariates

$$Y_i(v_i, v_{-i}) \rightarrow Y_i(v_i, v_i^S)$$

- Estimation of outcome model and propensity scores
 - Super-learner
 - Generalized linear model, elastic-net penalty, generalized additive model

Descriptive summary

		Overall (n=454)	Vaccinated (n=161)
Influenza-like illness		190 (42%)	65 (40%)
Lab-confirmed influenza		17 (4%)	2 (1%)
Female		269 (60%)	104 (65%)
Class year			
	Freshman	256 (57%)	97 (61%)
	Sophomore	111 (25%)	35 (22%)
	Junior	32 (7%)	12 (8%)
	Senior (4+)	50 (12%)	16 (10%)



Conclusions

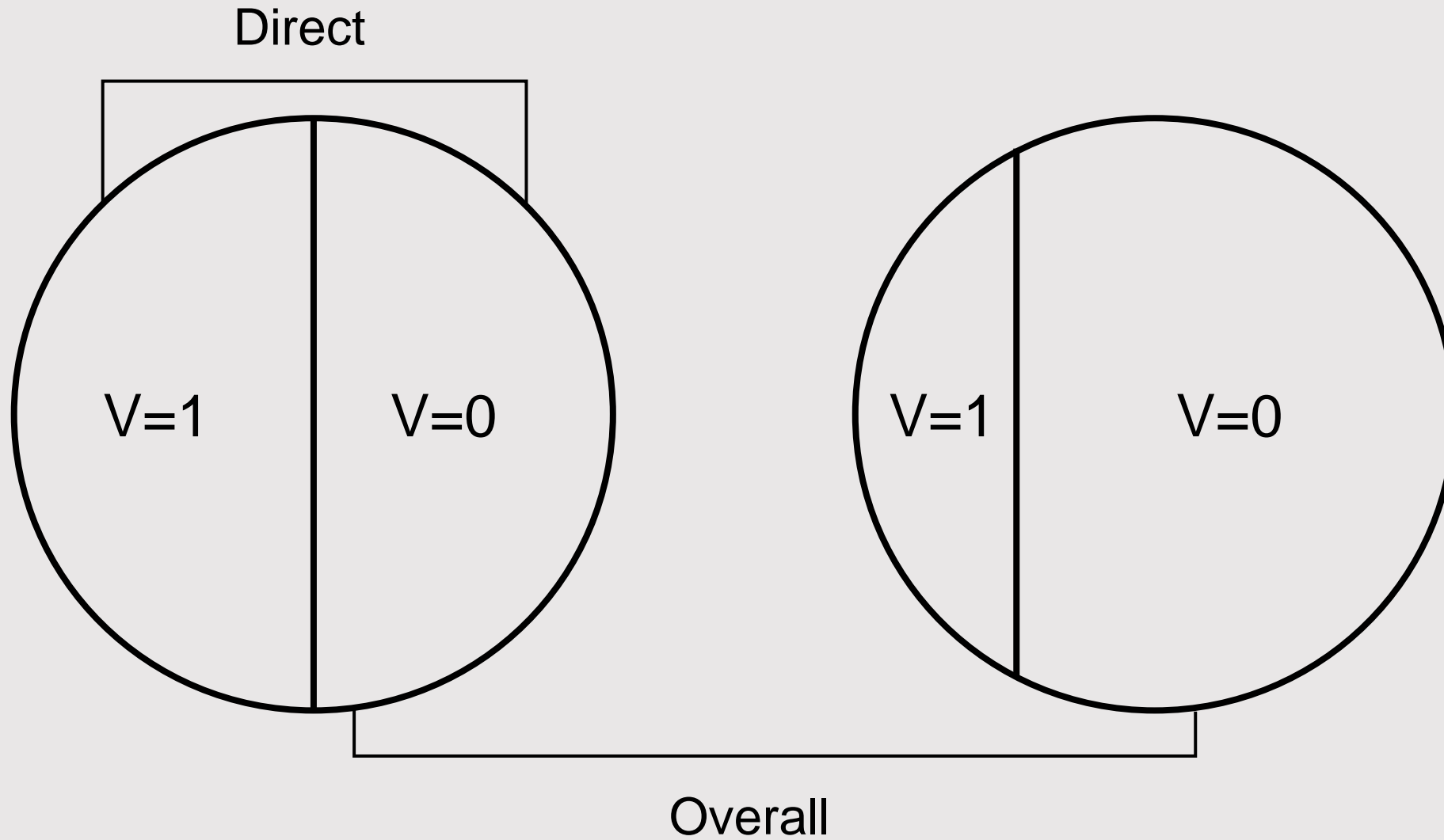
- General challenges to evaluating VE
 - Parameter of public health importance
 - COVID vaccination, masks, etc.
- Limitations
 - Power
 - Interference limited to immediate contacts

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Adapted from Halloran ME, Longini Jr IM, Struchiner CJ. Design and Analysis of Vaccine Studies. New York, NY: Springer; 2010