

# Precision of Estimates of Nonresponse Bias in Means

Stephanie Eckman  
Jennifer Unangst, Jill Dever, Chris Antoun



- Published estimates of NR bias
  - Judge quality of that survey
  - Predict nonresponse patterns in future surveys

Example: repeated cross-sectional survey

*Round X shows low response from HHs in Northwest region*

$$\text{bias}(p_r^{nw}) = p_r^{nw} - p_{full}^{nw}$$

*Should we change protocol in Round Y?*

- Maybe response is low because of *sample*
  - If we repeated survey, how would bias differ?



*sampling variability of  
NR bias estimates*

1. How variable are NR bias estimates?
  - RR
  - Clustering
2. How to estimate  $Var(bias(\bar{y}_r))$ ?



# Simulation Set Up: Unclustered

variable of interest  
~ response propensity

$$\begin{pmatrix} Y \\ Z \end{pmatrix} \sim N \left[ \begin{pmatrix} 10 \\ \delta \end{pmatrix}, \begin{pmatrix} \theta & \rho \\ \rho & 1 \end{pmatrix} \right]$$

---

## Parameters

---

$\delta$	RR	30%	70%
$\rho$	NR bias	None	Up to 10% NR bias

---

# Simulation Set Up: Clustered

variable of interest  
~ response propensity

$$\begin{pmatrix} Y \\ Z \end{pmatrix} \sim N \left[ \begin{pmatrix} 10 + \phi \\ \delta \end{pmatrix}, \begin{pmatrix} \theta & \rho \\ \rho & 1 \end{pmatrix} \right]$$

---

<b>Parameters</b>			
$\delta$	RR	30%	70%
$\rho$	NR bias	None	Up to 10% NR bias
$\phi$	Clustering	Unclustered	Clustered

---

# More Simulation Setup

- Used similar process to create binomial & Poisson  $Y$ s

- Correlated with  $Z$

- Normal  $\rightarrow$  uniform  $\rightarrow$  binomial

$$Y \sim \text{Bin}(1, 0.5)$$

$$Y \sim \text{Poisson}(3)$$

- 2,000 samples of  $n=1,000$

- SRS

- Multi-stage cluster

- Response propensity  $RP_k = \frac{e^{1+Z_k}}{1 + e^{1+Z_k}}$

deterministic:  $R_{det,k} = 0,1$

stochastic:  $R_{sto,s,k} = 0,1$

# Research Questions

1. How variable are NR bias estimates?
  - RR
  - Clustering
2. How to estimate  $Var(bias(\bar{y}_r))$ ?

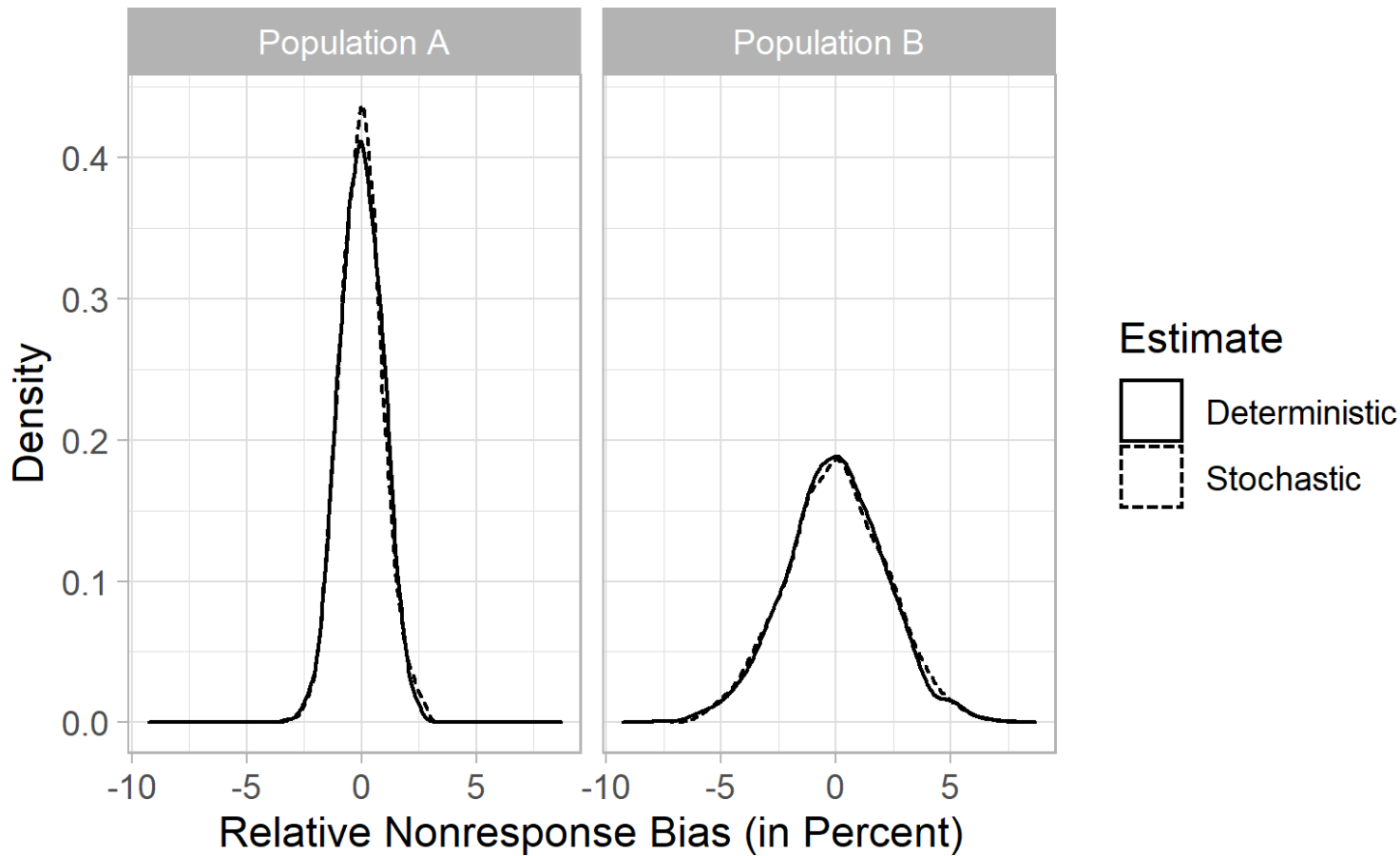




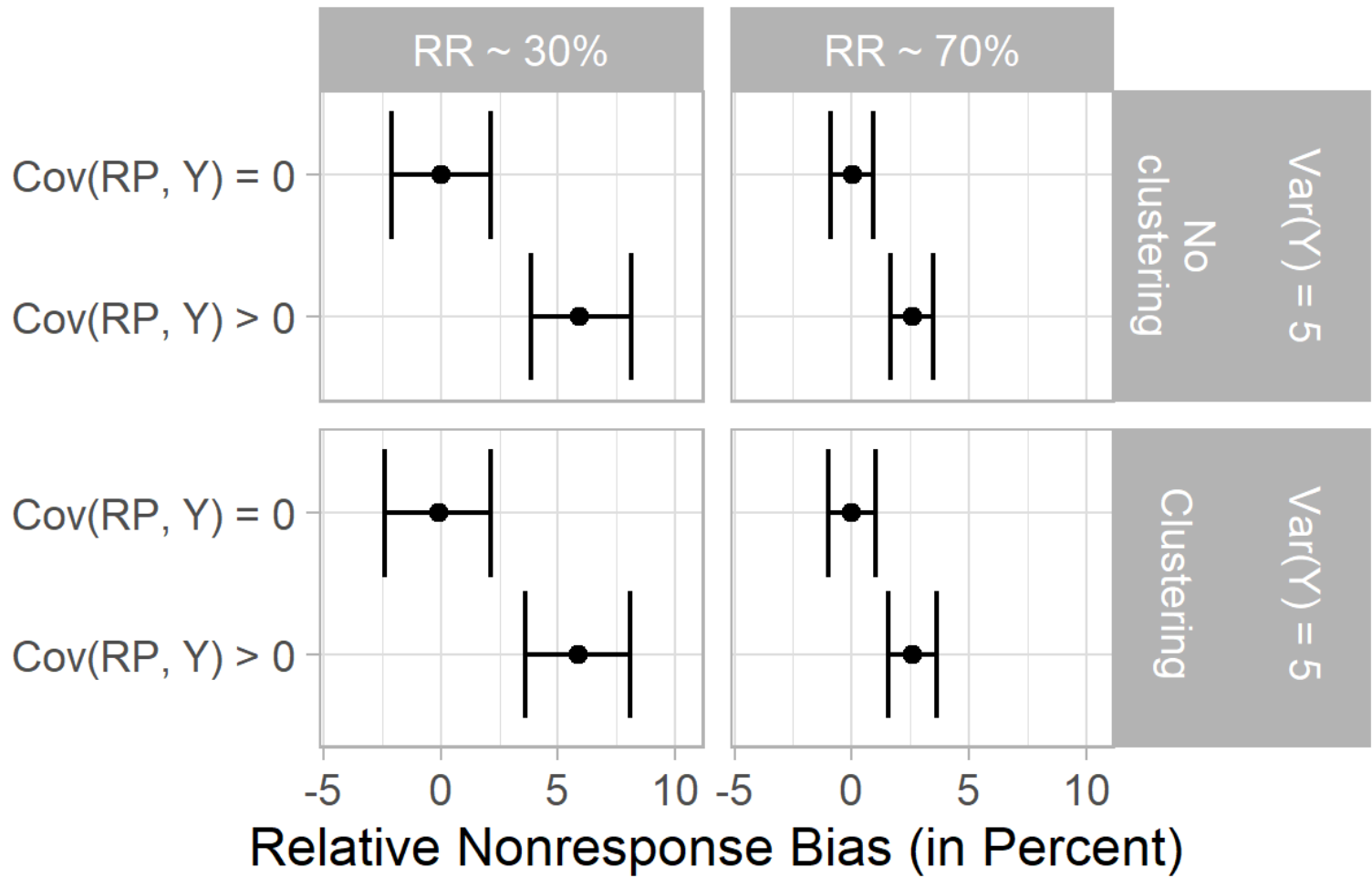
# Example Results: Estimated Bias in Simulated Samples

RR ~ 70%  
Bias = 0  
Unclustered

RR ~ 30%  
Bias = 0  
Clustered

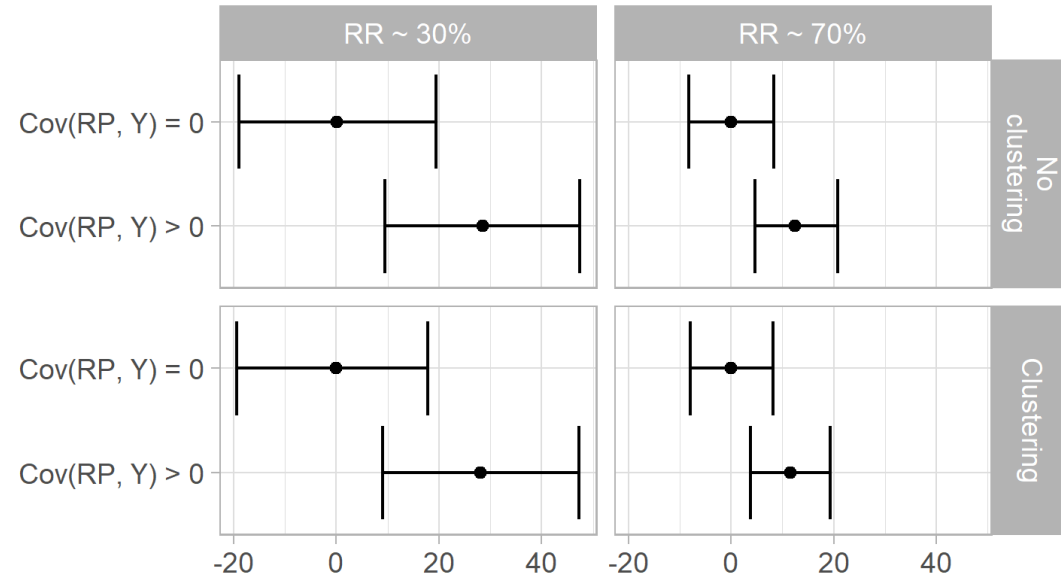


# All Populations: Estimated Bias in Simulated Samples

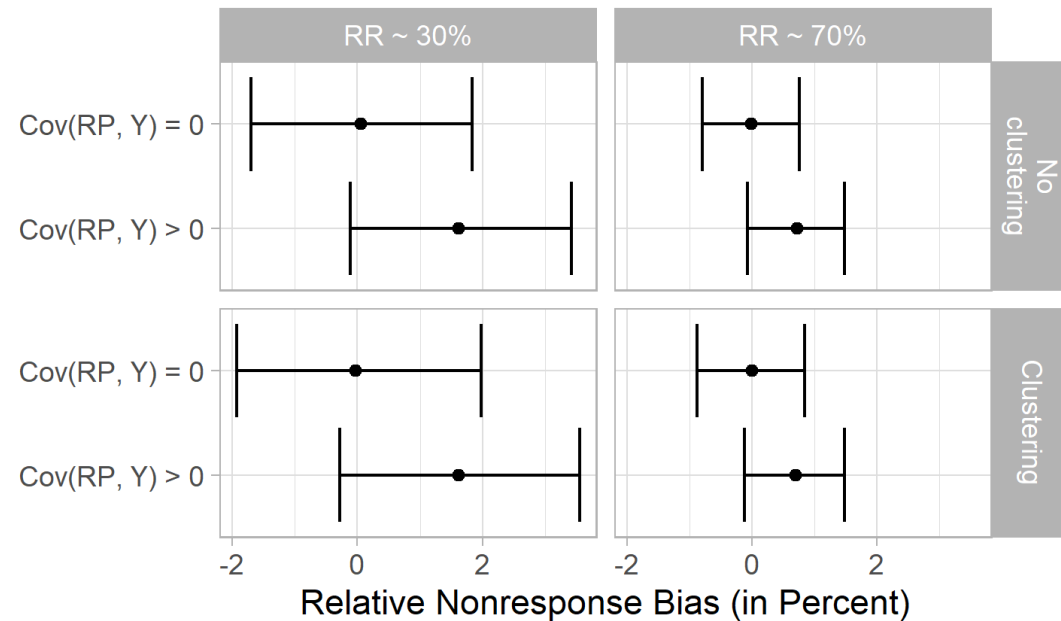


Graphs show 2.5% - 97.5% range

### Bernoulli, $p = 0.5$



### Poisson, $\omega = 3$



# Research Questions

1. How variable are NR bias estimates?
  - RR
  - Clustering
2. How to estimate  $Var(bias(\bar{y}_r))$ ?



- Lee (2006) method

$$\text{bias}(\bar{y}_r) = (1 - rr) \times (\bar{y}_r - \bar{y}_{nr})$$

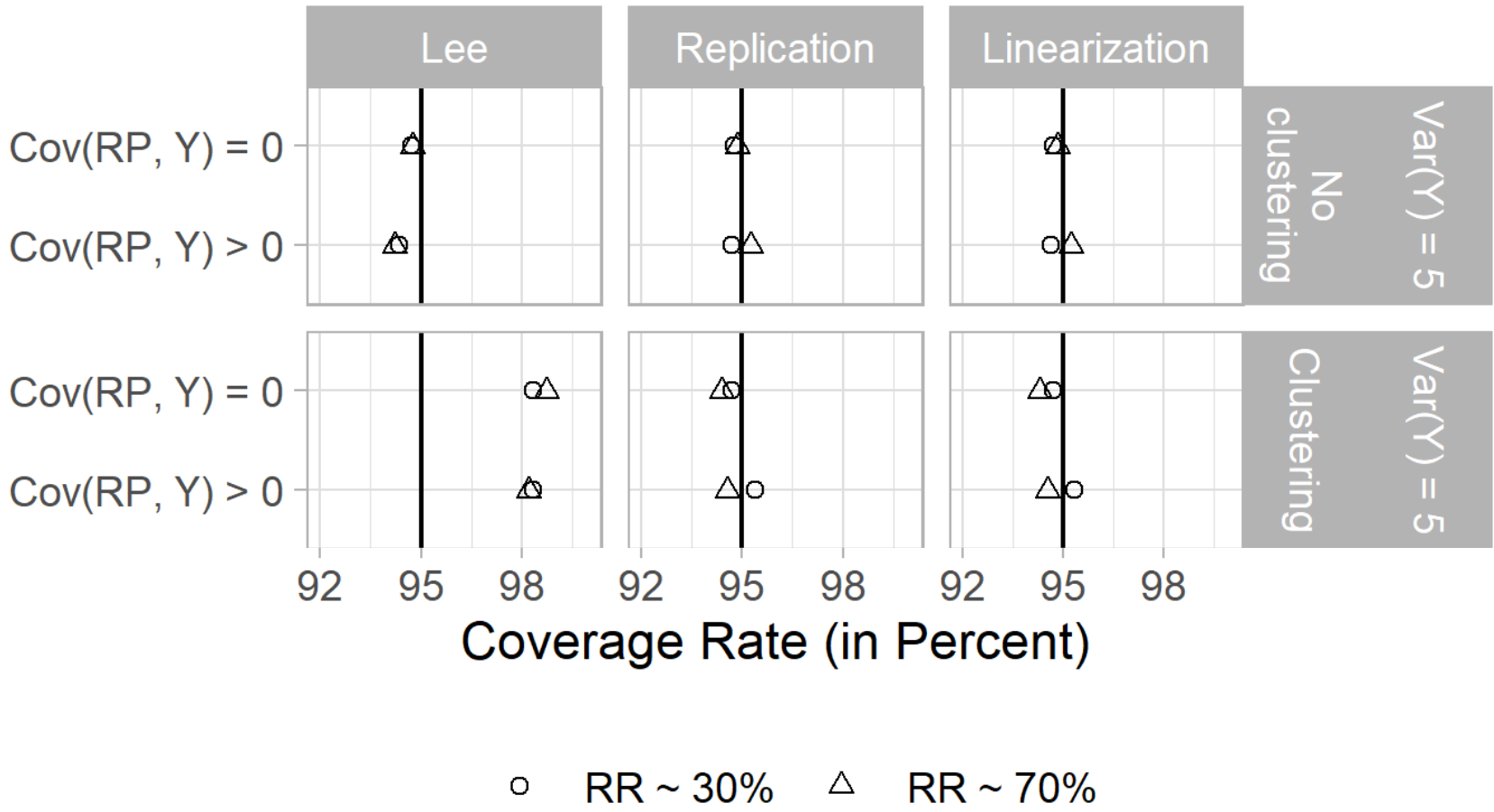
$$\text{Var}[\text{bias}(\bar{y}_r)] = (1 - rr)^2 \times [\text{Var}(\bar{y}_r) + \text{Var}(\bar{y}_{nr})]$$

# Variance Estimation Methods

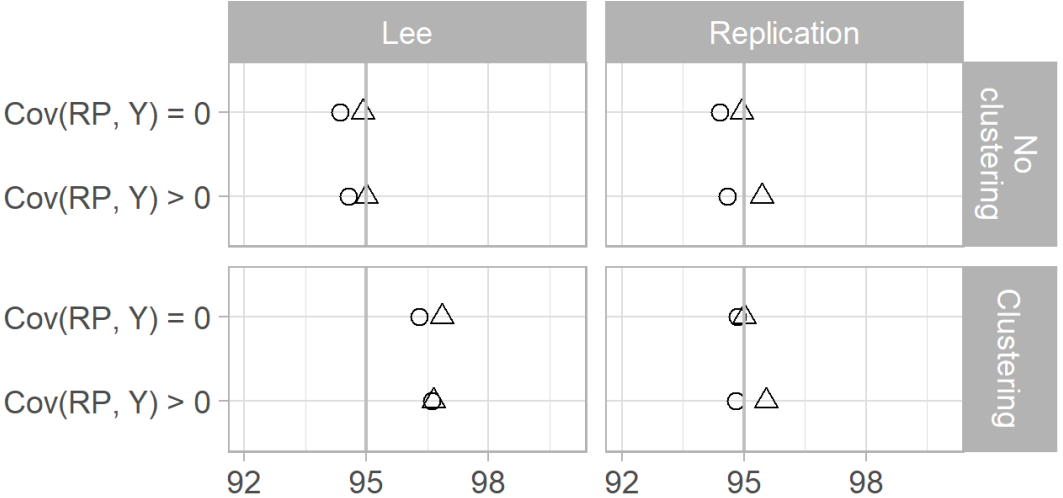
- Replication & Linearization

ID	Response?	Base weight	Adjusted weight	Y	Set	Analysis weight
1	Yes	100	140	10.3	1	140
2	Yes	150	190	9.8	1	190
3	No	80	0	8.7	1	0
1	Yes	100	140	10.3	2	100
2	Yes	150	190	9.8	2	150
3	No	80	0	8.7	2	80

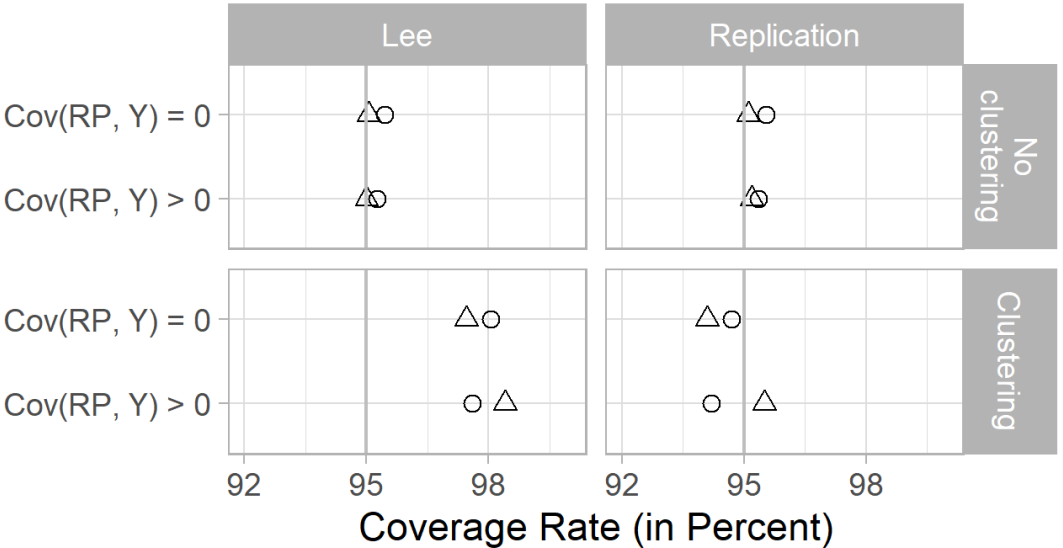
# Simulation Study – Performance of Three Variance Estimation Methods



### Binomial, $p = 0.5$



### Poisson, $\omega = 3$



○ RR ~ 30%   △ RR ~ 70%



# Application: LISS and GSS

Variable	Type	n	bias	Lee	Repl	Ratio
Hours of TV per weekday	Continuous	623	0.078	0.029	0.030	0.972
Ever driven drunk	Indicator	623	0.013	0.007	0.007	0.986
Immigrants improve NL	Indicator	622	0.011	0.007	0.007	0.987
Approval of Obama	Continuous	622	0.356	0.388	0.389	0.998
Times exercise per week	Count	623	0.003	0.008	0.008	1.004
Times restaurant last year	Count	623	-0.05	0.198	0.197	1.004
Occupational prestige	Continuous	1889	-0.012	0.187	0.204	0.918
Voted in 2004	Indicator	1754	0.023	0.006	0.003	1.725
Hours of TV per day	Continuous	1426	0	0.064	0.031	2.029
Finances better	Indicator	2033	-0.004	0.008	0.003	2.611
Believe in afterlife	Indicator	1803	0.017	0.009	0.003	2.895
Own gun	Indicator	1233	0	0.005	0.001	6.495

Ratio is Lee standard error / replication standard error

# Take Aways

- Estimates of NR bias in means vary across samples
- Lee method works well with unclustered samples
- Replication & linearization work with all samples
- Careful when generalizing from 1 sample to another



**Thank You**

**Stephanie Eckman**

Fellow, RTI International

Code, paper, slides:

<https://osf.io/rbzyd/>

[seckman@rti.org](mailto:seckman@rti.org)

<http://stepheckman.com>

@stephnie

	<b>Stata (linearization)</b>	<b>R (JK replication)</b>
Describe unclustered data	<code>svyset ID [pweight = v]</code>	<pre>dsg &lt;- assvrepdesign(   svydesign(ids = ~ID,   data = dset,   weights = ~v)</pre>
Describe clustered data	<code>svyset cluster [pweight = v]</code>	<pre>dsg &lt;- assvrepdesign(   svydesign(ids = ~cluster,   data = dset,   weights = ~v)</pre>
Estimate bias	<code>svy: mean y, over(set)</code>	<pre>bias &lt;- svyby(~y, ~set, dsg,   svymean,   covmat = TRUE)</pre>
Estimate standard error of bias estimate	<code>lincom _b[c.y@1bn.set] – _b[c.y@2.set]</code>	<pre>SE(svycontrast(bias,   quote('1' – '2')))</pre>