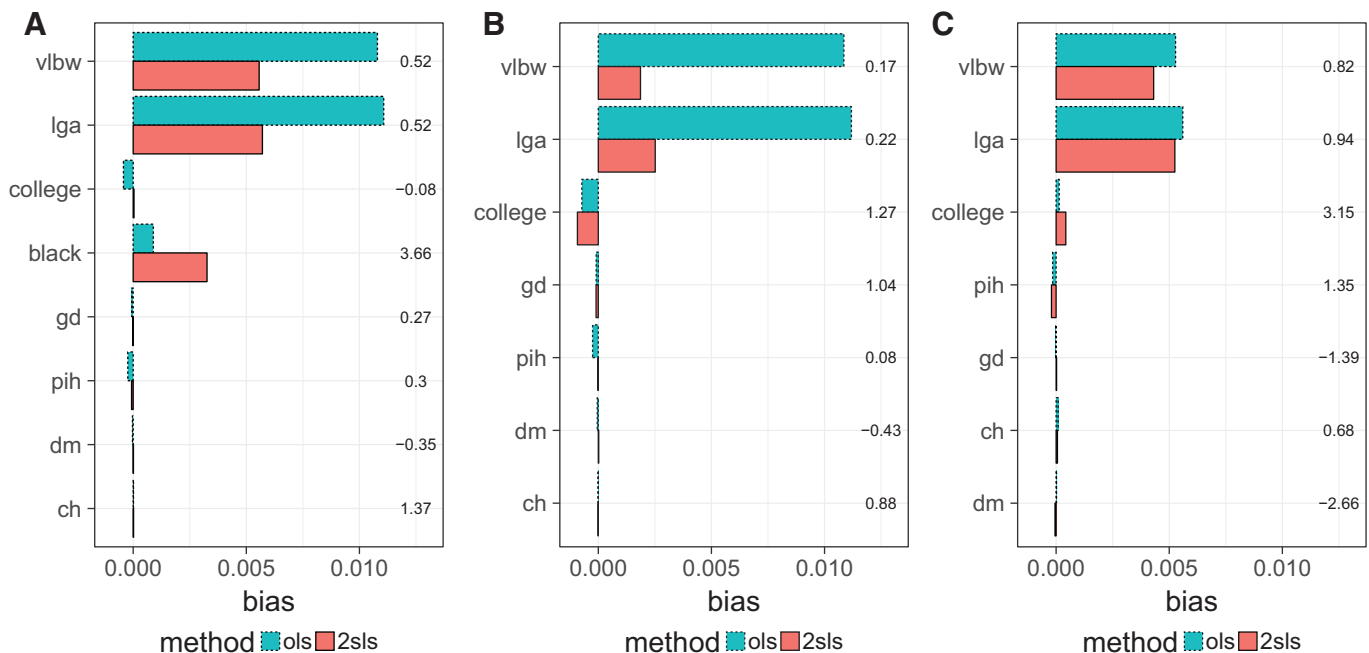


Upon restricting to 386,313 term births otherwise unaffected by an anomaly or obstetrical complication, the rate of maternal death within 24 hours was 3.0 versus 0.08 per 10,000 women, comparing Apgar scores of 0–6 versus 7–10—an aHR of 42 (4.3, 400; eFigure

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**FIGURE.** Diagnosis of IV (2SLS) bias compared with regression (OLS) bias from confounding. The bars show the absolute bias induced by a covariate that is computed by (1). The numbers on the right side of the bars are the bias ratios. Several covariates are recorded, including if the baby had vlbw, lga, mother had college degree (college), mother was African-American (black), mother had gd, dm, pih, or ch. A, Entire dataset. B, White mothers only. C, Black mothers only. ch indicates chronic hypertension; dm, diabetes mellitus; gd, gestational diabetes; lga, low gestational age; pih, pregnancy-induced hypertension; vlbw, very low birthweight.

founders. While this assumption cannot be verified empirically, its plausibility can be assessed by checking if the IV is independent of the measured confounders. If the IV is not independent of a measured confounder, it raises concerns when the measured confounder may be a proxy for other unmeasured confounders.<sup>4</sup>

In practice, a common approach is to compute the balance of observed covariates across levels of the IV and compare it with the balance across levels of the treatment.<sup>1</sup> This method is criticized (rightly, in our opinion) by Jackson and Swanson<sup>5</sup> as the bias of an IV estimate can be amplified by how strongly the IV is predictive of the treatment. Let  $D$  be the treatment,  $Y$  the outcome,  $Z$  the binary instrument,  $X$  an observed covariate, and  $Y^d$  the potential outcome for treatment level  $d$ . By assuming the linear structural model  $E[Y^d | X] = \alpha_0 + \alpha_1 d + \alpha_2 X$ , where the only confounder is  $X$ , Brookhart and Schneeweiss<sup>4</sup> derived the following bias formulas for two-stage least squares

(2SLS) IV and ordinary least squares regression (OLS) estimates that do not control for the measured covariate  $X$ :

$$\text{bias(2SLS estimator)} = \alpha_2 \cdot \frac{E[X | Z = 1] - E[X | Z = 0]}{E[D | Z = 1] - E[D | Z = 0]}, \quad (1)$$

$$\text{bias(OLS estimator)} = \alpha_2 \cdot (E[X | D = 1] - E[X | D = 0]).$$

Based on (1), Jackson and Swanson<sup>5</sup> proposed to assess the potential bias of an IV analysis by: (1) plotting the bias components in (1) for each covariate; (2) displaying a table of bias ratios.

We propose to augment Jackson and Swanson's<sup>5</sup> diagnostics with a plot that considers absolute bias. Large bias ratios can be misleading because not all covariates are confounders. Because the bias ratio of the two terms in (1) does not depend on  $\alpha_2$ , even if the bias ratio is large with respect to a covariate, the absolute bias may not be serious if  $\alpha_2$  is close to 0 ( $\alpha_2 = 0$  if  $X$  is not a confounder). Nevertheless, bias ratios can be more robust than absolute biases when the linear structural model is

misspecified. Our diagnostic barplots in the Figure show not only the bias ratios but also the absolute biases. The measured covariates are ordered from top to bottom by the bias amplifying factor  $|\alpha_2| \sqrt{\text{Var}(X)}$  (whose square is the variance of  $Y$  explained by  $X$ ), so that we might be most worried about high bias ratios in covariates nearer the top.

We demonstrate our proposal using the neonatal intensive care units (NICU) effectiveness study<sup>3,6</sup> in the Figure. In this study,  $D$  is an indicator of a premature baby being delivered in a high technology NICU and  $Y$  is in-hospital mortality. The IV  $Z = 1$ , if the excess travel time to the nearest high technology NICU is more than 10 minutes and  $Z = 0$  otherwise. When using the entire dataset (Figure A), it is evident that the IV analysis is helpful to reduce confounding bias due to health condition of the babies (vlbw and lga). Also, although the bias ratio with respect to chronic hypertension (ch) is larger than 1, this variable and the other measured maternal comorbidities are essentially uncorrelated with the outcome and the absolute confounding bias

is negligible. This suggests that bias from the IV being correlated with unmeasured maternal comorbidities might not be concerning. Race (black) has bias ratio larger than 1 and is predictive of the outcome. After conditioning on race (Figure B, C), it appears that the IV analysis might be less biased than regression for white mothers but comparably biased for black mothers.

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Validation of Severe Maternal Morbidity on the US Certificate of Live Birth

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To the Editor:

Severe maternal morbidity, unexpected outcomes of labor and delivery that have substantial short- or long-term health consequences, has been steadily increasing in recent years, affecting more than 50,000 women in the United States in 2013–2014.<sup>1</sup> It has been suggested that the rise is related to changes in the overall health of the child-bearing population, including increases in maternal age,<sup>2</sup> prepregnancy obesity,<sup>3</sup> preexisting chronic medical conditions,<sup>4</sup> and cesarean delivery.<sup>5</sup> The 2003 revision of the US Certificate of Live Birth added specific checkbox items on severe maternal morbidity occurring within 24 hours of delivery to facilitate national surveillance of these outcomes, as well as research on maternal complications of pregnancy. The items included maternal transfusion, third- or fourth-degree perineal laceration, ruptured uterus, unplanned hysterectomy, and admission to the intensive care unit. Massachusetts implemented the 2003 revision in 2011. As part of a larger study evaluating subfertility and maternal–child health, we evaluated the accuracy of severe maternal morbidity reported on the birth certificate, using hospital discharge data as the reference.

We designed a population-based study of all live births in Massachusetts between 2011 and 2013 linked to hospital discharge data. We used International Classification of Diseases, 9th revision (ICD-9) codes in the hospital discharge data to identify maternal blood transfusion (procedure codes 99.x), unplanned hysterectomy (procedure codes 68.3–68.9), ruptured uterus (diagnosis codes 665.0 and 665.1) and, among vaginal births, third- or fourth-degree perineal laceration (diagnosis code 664.2 and 664.3); intensive care unit (ICU) admission was identified by Uniform Billing (UB-04) revenue codes 0200–0209. The rates for each severe maternal morbidity measure were calculated per 100,000 deliveries. Sensitivity (probability of being included on the birth certificate when it was reported on the hospital discharge) and positive predictive value (PPV, probability of being on the hospital discharge when it was reported on the birth certificate), specificity, and negative predictive value were calculated to measure the reporting of severe maternal morbidity measures by mode of delivery (vaginal, cesarean), except for third- or fourth-degree perineal laceration (vaginal births only). We obtained institutional review board approval from the Massachusetts Department of Public Health,

Table. Reporting of Maternal Morbidity on the Birth Certificate Versus Hospital Discharge Massachusetts, 2011–2013

Maternal Morbidity	Mode of Delivery	Rate/100,000 Deliveries		Numerator Present in Both Sources	Sensitivity	Positive Predictive Value
		Birth Certificate	Hospital Discharge			
Blood transfusion	All	171	1053	261	0.12	0.73
	Vaginal	92	637	97	0.11	0.75
	Cesarean	343	1962	162	0.13	0.72
Ruptured uterus	All	102	71	39	0.26	0.18
	Vaginal	45	6	2	0.22	0.03
	Cesarean	229	213	37	0.26	0.25
Unplanned hysterectomy	All	58	75	80	0.51	0.66
	Vaginal	22	16	11	0.48	0.35
	Cesarean	137	203	69	0.52	0.77
Admission to intensive care	All	132	194	81	0.20	0.29
	Vaginal	92	60	18	0.21	0.14
	Cesarean	219	485	63	0.20	0.44
Third- or fourth- degree perineal laceration	Vaginal	928	2746	1190	0.30	0.90