

Supplementary material for “A goodness-of-fit test of logistic regression models for case-control data with measurement errors” by G. Xu and S. Wang

1. ADDITIONAL RESULTS FROM THE SIMULATION AND THE APPLICATION

Additional simulation results supplementary to Table 1 in the paper by Xu and Wang are given in Tables 1 and 2. Estimated coefficients in the Framingham heart disease data analysis and their standard errors are given in Table 3.

Table 1. Estimation accuracy in Simulation Examples 1 and 3.

$\theta = 0$	(n_1, n_2)	Method	X log-normal				X Y = 0 normal			
			$\hat{\beta}_Z$		$\hat{\beta}_X$		$\hat{\beta}_Z$		$\hat{\beta}_X$	
			Bias	RMSE	Bias	RMSE	Bias	RMSE	Bias	RMSE
	(100, 200)	M_n	0.06	0.25	-0.07	0.39	0.00	0.14	-0.03	0.19
		D_n	-0.24	0.30	0.47	0.49	0.00	0.14	0.18	0.22
		L_n	-0.24	0.30	0.47	0.49	0.00	0.14	0.18	0.22
	(200, 100)	M_n	0.03	0.26	-0.10	0.50	0.00	0.14	-0.03	0.19
		D_n	-0.25	0.29	0.51	0.52	0.00	0.14	0.18	0.23
		L_n	-0.25	0.29	0.51	0.52	0.00	0.14	0.18	0.23

M_n is the test statistic presented in this paper, D_n is the test statistic proposed by Zhang (2001) and L_n is the test statistic proposed by Lin et al. (2002). RMSE stands for root mean square error.

Table 2. Estimation accuracy in Simulation Example 2.

$\theta = 0$	(n_0, n_1)	Method	$\hat{\beta}'_Z$		$\hat{\beta}'_X$	
			Bias	RMSE	Bias	RMSE
	(100, 200)	D_n	0.003	0.17	-0.01	0.19
	(200, 100)	D_n	0.010	0.17	-0.01	0.21

RMSE stands for root mean square error. The true values of coefficients $\beta'_Z = 0.4$ and $\beta'_X = -0.9$.

Table 3. Estimated coefficients in the Framingham heart disease data analysis

	Method	Age	Smoke	Log-Chol	Log-SBP
Estimates	M_n	0.052	0.53	4.02	3.07
Estimates	D_n	0.056	0.53	2.79	2.29
SE	M_n	0.018	0.37	1.19	0.81
SE	D_n	0.017	0.35	0.78	0.58

M_n : our method. D_n : method of Zhang (2001). SE: standard error.

2. ADDITIONAL PROOFS

Claim 1: Under error structure (5), model (1) leads to model (6).

49 *Proof.* If model (1) is true, then

$$\begin{aligned} 50 \quad \text{pr}(Y = y | Z, X, \Delta) &\propto \text{pr}(Y = y | Z, X) f(\Delta | Y = y, X) \\ 51 &\propto \exp\{y(\beta_0^* + Z^T \beta_Z + \Delta^T \beta_X - \beta_X^T \Sigma_U \beta_X / 2)\}, \end{aligned}$$

52 which does not depend on X , where $f(\Delta | Y = y, X)$ is the density function of $\Delta | Y = y, X$.
53 Similar definitions are used below. Hence

$$\frac{\text{pr}(Y = 1 | Z, \Delta)}{\text{pr}(Y = 0 | Z, \Delta)} = \exp\{\beta_0^* + Z^T \beta_Z + \Delta^T \beta_X - \beta_X^T \Sigma_U \beta_X / 2\},$$

54 which further implies that

$$\frac{g_1(Z, \Delta)}{g_0(Z, \Delta)} = \exp\{\beta_0 + Z^T \beta_Z + \Delta^T \beta_X - \beta_X^T \Sigma_U \beta_X / 2\}.$$

55 Therefore, model (1) leads to model (6) under error structure (5). □

56 *Proof of Lemma 2.* If model (8) is true, then

$$\begin{aligned} 57 \quad \text{pr}(Y = y | Z, X, \Delta) \\ 58 &= \{r^*(Z, X, \eta)\}^y \frac{\exp\{y(\beta_0^* + Z^T \beta_Z + \Delta^T \beta_X - \beta_X^T \Sigma_U \beta_X / 2)\}}{f(\Delta | Z, X) \{1 + r^*(Z, X, \eta) \exp(\beta_0^* + Z^T \beta_Z + X^T \beta_X)\}}. \end{aligned}$$

59 Let $r_1^*(Z, X, \Delta, \eta) = f(\Delta | Z, X) \{1 + r^*(Z, X, \eta) \exp(\beta_0^* + Z^T \beta_Z + X^T \beta_X)\}$ and
60 $r_2^*(Z, X, \Delta, \eta) = r^*(Z, X, \eta) / r_1^*(Z, X, \Delta, \eta)$. Then

$$\text{pr}(Y = 1 | Z, X, \Delta) = r_2^*(Z, X, \Delta, \eta) \exp(\beta_0^* + Z^T \beta_Z + \Delta^T \beta_X - \beta_X^T \Sigma_U \beta_X / 2)$$

61 and

$$\text{pr}(Y = 0 | Z, X, \Delta) = 1 / r_1^*(Z, X, \Delta, \eta).$$

62 Thus,

$$\frac{\text{pr}(Y = 1 | Z, \Delta)}{\text{pr}(Y = 0 | Z, \Delta)} = r(Z, \Delta, \eta) \exp(\beta_0^* + Z^T \beta_Z + \Delta^T \beta_X - \beta_X^T \Sigma_U \beta_X / 2),$$

63 where $r(Z, \Delta, \eta) = \int_X r_2^*(Z, x, \Delta, \eta) f_X(x) dx / \int_X \{r_1^*(Z, x, \Delta, \eta)\}^{-1} f_X(x) dx$. Since for all
64 (Z, X) , $r^*(Z, X, \eta_0) = 1$, one has $r(Z, \Delta, \eta_0) = 1$ for all (Z, Δ) . Straightforward algebra shows
65 that

$$\frac{g_1(Z, \Delta)}{g_0(Z, \Delta)} = r(Z, \Delta, \eta) \exp(\beta_0 + Z^T \beta_Z + \Delta^T \beta_X - \beta_X^T \Sigma_U \beta_X / 2),$$

66 which completes the proof. □

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