## Supplement to "Tests for Gene-Environment Interactions and Joint Effects with Exposure Misclassification"

Running head: GxE Interactions with Exposure Misclassification

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## Web Appendix 1

In the following algebraic development, we develop exact expressions for the log-odds ratios  $\beta_E$ ,  $\beta_G$ , and  $\beta_{GE}$  as functions of the quantities  $\alpha_G$ ,  $\alpha_E$ ,  $\theta_{GE}$ ,  $P_G \equiv \Pr(G = 1|D = 0)$ , and  $P_E \equiv \Pr(E = 1|D = 0)$ . As given in the text, the control probabilities relate to  $\theta_{GE}$ ,  $P_G$ , and  $P_E$  according to

$$\exp\{\theta_{GE}\} = \frac{p_{000}(p_{000} - (1 - P_G - P_E))}{(1 - P_G - p_{000})(1 - P_E - p_{000})},$$
$$p_{001} = 1 - P_G - p_{000}, \ p_{010} = 1 - P_E - p_{000}$$

The case probabilities are then given by  $p_{100} \propto p_{000}$ ,  $p_{101} \propto \exp{\{\beta_E\}}p_{001}$ ,  $p_{110} \propto \exp{\{\beta_G\}}p_{010}$ , and  $p_{111} \propto \exp{\{\beta_E + \beta_G + \beta_{GE}\}}p_{011}$ , normalized to sum to one. Thus, the marginal log-ORs,  $\alpha_G$ and  $\alpha_E$ , are written as

$$\begin{aligned} \alpha_{G} &= \log\left(\frac{p_{111} + p_{110}}{p_{101} + p_{100}}\right) + \log\left(\frac{p_{001} + p_{000}}{p_{011} + p_{010}}\right) \\ &= \log\left(\frac{\exp\{\beta_{E} + \beta_{G} + \beta_{GE}\}p_{011} + \exp\{\beta_{G}\}p_{010}}{\exp\{\beta_{E}\}p_{001} + p_{000}}\right) + \log\left(\frac{p_{001} + p_{000}}{p_{011} + p_{010}}\right) \\ &= \beta_{G} + \log\left(\frac{\exp\{\beta_{E} + \beta_{GE}\}p_{011} + p_{010}}{\exp\{\beta_{E}\}p_{001} + p_{000}}\right) + \log\left(\frac{p_{001} + p_{000}}{p_{011} + p_{010}}\right) \end{aligned}$$
(W1)  
$$\alpha_{E} &= \log\left(\frac{p_{111} + p_{101}}{p_{110} + p_{100}}\right) + \log\left(\frac{p_{010} + p_{000}}{p_{011} + p_{001}}\right) \\ &= \log\left(\frac{\exp\{\beta_{E} + \beta_{G} + \beta_{GE}\}p_{011} + \exp\{\beta_{E}\}p_{001}}{\exp\{\beta_{G}\}p_{010} + p_{000}}\right) + \log\left(\frac{p_{010} + p_{000}}{p_{011} + p_{001}}\right) \\ &= \beta_{E} + \log\left(\frac{\exp\{\beta_{G} + \beta_{GE}\}p_{011} + p_{001}}{\exp\{\beta_{G}\}p_{010} + p_{000}}\right) + \log\left(\frac{p_{010} + p_{000}}{p_{011} + p_{001}}\right), \end{aligned}$$
(W2)

Thus, the marginal log-ORs  $\alpha_G$  and  $\alpha_E$  can be written as functions of the control probability vector and the ORs  $\beta_G$ ,  $\beta_E$ , and  $\beta_{GE}$ , and specification of any three of  $\alpha_G$ ,  $\alpha_E$ ,  $\beta_G$ ,  $\beta_E$ , or  $\beta_{GE}$  determine the value of the remaining two.

Web Table 1: Simulation settings for additional GEI results, given in Web Figures 1–6 (top), and additional gene discovery results, given in Web Figures 7–9 (bottom)<sup>a</sup>

Web Figure	range $\exp\{\beta_{GE}\}$	$\beta_G$	$P_E$	$\alpha_E$	$n_0, n_1$	$p_{ind}$	$\#\{\beta_G^{\text{NULL}} \neq 0\}$
1	(1.00, 1.75)	$\log(1.2)$	0.3	$\log(1.5)$	$2 \times 10^4$	0.995	0
2	(1.00, 1.35)	$\log(1.0)$	0.3	$\log(1.5)$	$2 \times 10^4$	0.995	0
3	(1.00, 1.35)	$\log(1.2)$	0.3	$\log(1.5)$	$10^{4}$	0.995	0
4	(1.00, 1.35)	$\log(1.2)$	0.1	$\log(1.75)$	$2 \times 10^4$	0.995	0
5	(1.00, 1.75)	$\log(1.2)$	0.1	$\log(1.75)$	$2 \times 10^4$	0.995	0
6	(1.00, 1.35)	$\log(1.2)$	0.3	$\log(1.5)$	$2 \times 10^4$	0.995	500
7	(1.00, 1.75)	$\log(1.0)$	0.3	$\log(1.5)$	$2 \times 10^4$	_	_
8	(1.00, 1.35)	$\log(1.2)$	0.3	$\log(1.5)$	$2 \times 10^4$	_	—
9	(1.00, 1.75)	$\log(1.0)$	0.1	$\log(1.75)$	$2 \times 10^4$	—	_

Abbreviations: GEI, gene-environment interaction.

<sup>*a*</sup> Those items in red indicate differences in settings from Figure 1 (GEI) or Figure 2 (gene discovery) in the main text. In regards to the last column, this gives the number of null markers, i.e.  $\beta_{GE} = 0$ , with genetic main effects sampled from  $\beta_G \sim \text{Unif}(\log(1.05), \log(1.2))$ . Each gene discovery method tests each marker independently. Thus, because we focus only on markers for which  $\beta_{GE} \neq 0$ , we do not need to consider parameters whose scope is limited to null markers, i.e.  $p_{\text{ind}}$  and  $\#\{\beta_G^{\text{NULL}} \neq 0\}$ .



Web Figure 1: Empirical power to detect gene-environment interaction in one marker for 7 GEI methods (CC, case-control; CO, case-only; EB, empirical Bayes; TS, two-step gene-environment screening; H2, hybrid two-step; EDGxE, joint marginal/association screening; CT, cocktail) and the marginal (MA) method from 5,000 datasets with n = 20,000 each of cases and controls and M = 100,000 - 1 null genetic markers. From top to bottom, each row corresponds to perfect classification, non-differential misclassification (sensitivity and specificity of 0.8), and differential misclassification (sensitivity of 1 and specificity of 0.8 for cases and sensitivity and specificity of 0.8 for controls) of the exposure variable. From left to right, each column corresponds to  $\theta_{GE} = \log(0.8), \theta_{GE} = 0,$  and  $\theta_{GE} = \log(1.1)$ . The exposure prevalence was  $P_E = 0.3$  and the marginal exposure log-OR was  $\alpha_E = \log(1.5)$ . For the non-null marker, the main genetic log-OR was  $\beta_G = \log(1.2)$  and the carrier prevalence was  $P_G = 0.36$ . For each null marker,  $\beta_G = 0$  and  $P_G = f^2 + 2f(1 - f)$ , where  $f \sim \text{Unif}[0.1, 0.3]$  is the minor allele frequency. These settings are identical to those of Figure 1 in the main text, but the range of  $\exp\{\beta_{GE}\}$  extends to 1.75



Web Figure 2: Empirical power to detect gene-environment interaction in one marker for 7 GEI methods (CC, case-control; CO, case-only; EB, empirical Bayes; TS, two-step gene-environment screening; H2, hybrid two-step; EDGxE, joint marginal/association screening; CT, cocktail) and the marginal (MA) method from 5,000 datasets with n = 20,000 each of cases and controls and M = 100,000 - 1 null genetic markers. From top to bottom, each row corresponds to perfect classification, non-differential misclassification (sensitivity and specificity of 0.8), and differential misclassification (sensitivity of 1 and specificity of 0.8 for cases and sensitivity and specificity of 0.8 for controls) of the exposure variable. From left to right, each column corresponds to  $\theta_{GE} = \log(0.8)$ ,  $\theta_{GE} = 0$ , and  $\theta_{GE} = \log(1.1)$ . The exposure prevalence was  $P_E = 0.3$  and the marginal exposure log-OR was  $\alpha_E = \log(1.5)$ . For the non-null marker, the main genetic log-OR was  $\beta_G = 0$  and the carrier prevalence was  $P_G = 0.36$ . For each null marker,  $\beta_G = 0$  and  $P_G = f^2 + 2f(1 - f)$ , where  $f \sim \text{Unif}[0.1, 0.3]$  is the minor allele frequency.



Web Figure 3: Empirical power to detect gene-environment interaction in one marker for 7 GEI methods (CC, case-control; CO, case-only; EB, empirical Bayes; TS, two-step gene-environment screening; H2, hybrid two-step; EDGxE, joint marginal/association screening; CT, cocktail) and the marginal (MA) method from 5,000 datasets with n = 10,000 each of cases and controls and M = 100,000 - 1 null genetic markers. From top to bottom, each row corresponds to perfect classification, non-differential misclassification (sensitivity and specificity of 0.8), and differential misclassification (sensitivity of 0.8 for cases and sensitivity and specificity of 0.8 for controls) of the exposure variable. From left to right, each column corresponds to  $\theta_{GE} = \log(0.8)$ ,  $\theta_{GE} = 0$ , and  $\theta_{GE} = \log(1.1)$ . The exposure prevalence was  $P_E = 0.3$  and the marginal exposure log-OR was  $\alpha_E = \log(1.5)$ . For the non-null marker, the main genetic log-OR was  $\beta_G = \log(1.2)$  and the carrier prevalence was  $P_G = 0.36$ . For each null marker,  $\beta_G = 0$  and  $P_G = f^2 + 2f(1 - f)$ , where  $f \sim \text{Unif}[0.1, 0.3]$  is the minor allele frequency.



Web Figure 4: Empirical power to detect gene-environment interaction in one marker for 7 GEI methods (CC, case-control; CO, case-only; EB, empirical Bayes; TS, two-step gene-environment screening; H2, hybrid two-step; EDGxE, joint marginal/association screening; CT, cocktail) and the marginal (MA) method from 5,000 datasets with n = 20,000 each of cases and controls and M = 100,000 - 1 null genetic markers. From top to bottom, each row corresponds to perfect classification, non-differential misclassification (sensitivity and specificity of 0.8), and differential misclassification (sensitivity of 0.8 for cases and sensitivity and specificity of 0.8 for controls) of the exposure variable. From left to right, each column corresponds to  $\theta_{GE} = \log(0.8), \theta_{GE} = 0, \text{ and } \theta_{GE} = \log(1.1)$ . The exposure prevalence was  $P_E = 0.1$  and the marginal exposure log-OR was  $\alpha_E = \log(1.75)$ . For the non-null marker, the main genetic log-OR was  $\beta_G = \log(1.2)$  and the carrier prevalence was  $P_G = 0.36$ . For each null marker,  $\beta_G = 0$  and  $P_G = f^2 + 2f(1 - f)$ , where  $f \sim \text{Unif}[0.1, 0.3]$  is the minor allele frequency.



Web Figure 5: Empirical power to detect gene-environment interaction in one marker for 7 GEI methods (CC, case-control; CO, case-only; EB, empirical Bayes; TS, two-step gene-environment screening; H2, hybrid two-step; EDGxE, joint marginal/association screening; CT, cocktail) and the marginal (MA) method from 5,000 datasets with n = 20,000 each of cases and controls and M = 100,000 - 1 null genetic markers. From top to bottom, each row corresponds to perfect classification, non-differential misclassification (sensitivity and specificity of 0.8), and differential misclassification (sensitivity of 1 and specificity of 0.8 for cases and sensitivity and specificity of 0.8 for controls) of the exposure variable. From left to right, each column corresponds to  $\theta_{GE} = \log(0.8), \theta_{GE} = 0, \text{ and } \theta_{GE} = \log(1.1)$ . The exposure prevalence was  $P_E = 0.1$  and the marginal exposure log-OR was  $\alpha_E = \log(1.75)$ . For the non-null marker, the main genetic log-OR was  $\beta_G = \log(1.2)$  and the carrier prevalence was  $P_G = 0.36$ . For each null marker,  $\beta_G = 0$  and  $P_G = f^2 + 2f(1 - f)$ , where  $f \sim \text{Unif}[0.1, 0.3]$  is the minor allele frequency. These settings are identical to those of Figure 4, but the range of exp{ $\beta_{GE}$ } extends to 1.75.



Web Figure 6: Empirical power to detect gene-environment interaction in one marker for 7 GEI methods (CC, case-control; CO, case-only; EB, empirical Bayes; TS, two-step gene-environment screening; H2, hybrid two-step; EDGxE, joint marginal/association screening; CT, cocktail) and the marginal (MA) method from 5,000 datasets with n = 20,000 each of cases and controls and M = 100,000 - 1 null genetic markers. From top to bottom, each row corresponds to perfect classification, non-differential misclassification (sensitivity and specificity of 0.8), and differential misclassification (sensitivity of 1 and specificity of 0.8 for cases and sensitivity and specificity of 0.8 for controls) of the exposure variable. From left to right, each column corresponds to  $\theta_{GE} = \log(0.8), \theta_{GE} = 0$ , and  $\theta_{GE} = \log(1.1)$ . The exposure prevalence was  $P_E = 0.3$  and the marginal exposure log-OR was  $\alpha_E = \log(1.5)$ . For the non-null marker, the main genetic log-OR was  $\beta_G = \log(1.2)$  and the carrier prevalence was  $P_G = 0.36$ . For 500 null markers,  $\beta_G \sim \text{Unif}[\log(1.05), \log(1.2)]$ , with  $\beta_G = 0$  for the remainder. For all null markers,  $P_G = f^2 + 2f(1 - f)$ , where  $f \sim \text{Unif}[0.1, 0.3]$  is the minor allele frequency.



Web Figure 7: Empirical power for discovery of one marker for the case-control method (CC) and 7 gene discovery methods (MA, marginal; JOINT(CC), 2-DF joint test; JOINT(EB), empirical Bayes 2-DF joint test; MA+CC, marginal + CC; MA+EB, marginal + empirical Bayes; CC(EXP), CC applied to exposed subgroup; EB(EXP), empirical Bayes applied to exposed subgroup) from 5,000 datasets with n = 20,000 each of cases and controls. From top to bottom, each row corresponds to perfect classification, non-differential misclassification (sensitivity and specificity of 0.8), and differential misclassification (sensitivity of 1 and specificity of 0.8 for cases and sensitivity and specificity of 0.8 for controls) of the exposure variable. From left to right, each column corresponds to  $\theta_{GE} = \log(0.8)$ ,  $\theta_{GE} = 0$ , and  $\theta_{GE} = \log(1.1)$ . The exposure prevalence was  $P_E = 0.3$  and the marginal exposure log-OR was  $\alpha_E = \log(1.5)$ . The main genetic log-OR was  $\beta_G = 0$  and the carrier prevalence was  $P_G = 0.36$ . These settings are identical to those of Figure 2 in the main text, but the range of  $\exp\{\beta_{GE}\}$  extends to 1.75.



Web Figure 8: Empirical power for discovery of one marker for the case-control method (CC) and 7 gene discovery methods (MA, marginal; JOINT(CC), 2-DF joint test; JOINT(EB), empirical Bayes 2-DF joint test; MA+CC, marginal + CC; MA+EB, marginal + empirical Bayes; CC(EXP), CC applied to exposed subgroup; EB(EXP), empirical Bayes applied to exposed subgroup) from 5,000 datasets with n = 20,000 each of cases and controls. From top to bottom, each row corresponds to perfect classification, non-differential misclassification (sensitivity and specificity of 0.8), and differential misclassification (sensitivity of 1 and specificity of 0.8 for cases and sensitivity and specificity of 0.8 for controls) of the exposure variable. From left to right, each column corresponds to  $\theta_{GE} = \log(0.8)$ ,  $\theta_{GE} = 0$ , and  $\theta_{GE} = \log(1.1)$ . The exposure prevalence was  $P_E = 0.3$  and the marginal exposure log-OR was  $\alpha_E = \log(1.5)$ . The main genetic log-OR was  $\beta_G = \log(1.2)$  and the carrier prevalence was  $P_G = 0.36$ .



Web Figure 9: Empirical power for discovery of one marker for the case-control method (CC) and 7 gene discovery methods (MA, marginal; JOINT(CC), 2-DF joint test; JOINT(EB), empirical Bayes 2-DF joint test; MA+CC, marginal + CC; MA+EB, marginal + empirical Bayes; CC(EXP), CC applied to exposed subgroup; EB(EXP), empirical Bayes applied to exposed subgroup) from 5,000 datasets with n = 20,000 each of cases and controls. From top to bottom, each row corresponds to perfect classification, non-differential misclassification (sensitivity and specificity of 0.8), and differential misclassification (sensitivity of 1 and specificity of 0.8 for cases and sensitivity and specificity of 0.8 for controls) of the exposure variable. From left to right, each column corresponds to  $\theta_{GE} = \log(0.8)$ ,  $\theta_{GE} = 0$ , and  $\theta_{GE} = \log(1.1)$ . The exposure prevalence was  $P_E = 0.1$  and the marginal exposure log-OR was  $\alpha_E = \log(1.75)$ . The main genetic log-OR was  $\beta_G = 0$  and the carrier prevalence was  $P_G = 0.36$ .