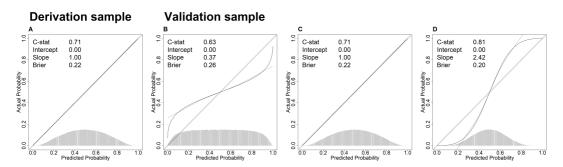
## Random measurement heterogeneity



**FIGURE 5** Predictive performance of a single-predictor binary logistic regression model. The predictor measurement structure corresponds to:

 $\begin{array}{lll} \text{A. } W_D = X + \varepsilon_D, & \text{where } X \sim \mathcal{N}(0,0.5) \text{ and } \varepsilon_D \sim \mathcal{N}(0,0.5). \\ \text{B. } W_V = X + \varepsilon_V, & \text{where } X \sim \mathcal{N}(0,0.5) \text{ and } \varepsilon_V \sim \mathcal{N}(0,2.0). \\ \text{C. } W_V = X + \varepsilon_V, & \text{where } X \sim \mathcal{N}(0,0.5) \text{ and } \varepsilon_V \sim \mathcal{N}(0,0.5). \\ \text{D. } W_V = X, & \text{where } X \sim \mathcal{N}(0,0.5). \\ \end{array} \qquad \begin{array}{ll} \text{Measurements are less precise at validation, i.e. } \sigma_{\varepsilon(D)}^2 < \sigma_{\varepsilon(V)}^2. \\ \text{Measurements are more precise at validation, i.e. } \sigma_{\varepsilon(D)}^2 > \sigma_{\varepsilon(V)}^2. \\ \text{Measurements are more precise at validation, i.e. } \sigma_{\varepsilon(D)}^2 > \sigma_{\varepsilon(V)}^2. \\ \text{Measurements are more precise at validation, i.e. } \sigma_{\varepsilon(D)}^2 > \sigma_{\varepsilon(V)}^2. \\ \text{Measurements are more precise at validation, i.e. } \sigma_{\varepsilon(D)}^2 > \sigma_{\varepsilon(V)}^2. \\ \text{Measurements are more precise at validation, i.e. } \sigma_{\varepsilon(D)}^2 > \sigma_{\varepsilon(V)}^2. \\ \text{Measurements are more precise at validation, i.e. } \sigma_{\varepsilon(D)}^2 > \sigma_{\varepsilon(V)}^2. \\ \text{Measurements are more precise at validation, i.e. } \sigma_{\varepsilon(D)}^2 > \sigma_{\varepsilon(V)}^2. \\ \text{Measurements are more precise at validation, i.e. } \sigma_{\varepsilon(D)}^2 > \sigma_{\varepsilon(V)}^2. \\ \text{Measurements are more precise at validation, i.e. } \sigma_{\varepsilon(D)}^2 > \sigma_{\varepsilon(V)}^2. \\ \text{Measurements are more precise at validation, i.e. } \sigma_{\varepsilon(D)}^2 > \sigma_{\varepsilon(V)}^2. \\ \text{Measurements are more precise at validation, i.e. } \sigma_{\varepsilon(D)}^2 > \sigma_{\varepsilon(V)}^2. \\ \text{Measurements are more precise at validation, i.e. } \sigma_{\varepsilon(D)}^2 > \sigma_{\varepsilon(V)}^2. \\ \text{Measurements are more precise at validation, i.e. } \sigma_{\varepsilon(D)}^2 > \sigma_{\varepsilon(V)}^2. \\ \text{Measurements are more precise at validation, i.e. } \sigma_{\varepsilon(D)}^2 > \sigma_{\varepsilon(V)}^2. \\ \text{Measurements are more precise at validation, i.e. } \sigma_{\varepsilon(D)}^2 > \sigma_{\varepsilon(V)}^2. \\ \text{Measurements are more precise at validation, i.e. } \sigma_{\varepsilon(D)}^2 > \sigma_{\varepsilon(V)}^2. \\ \text{Measurements are more precise at validation, i.e. } \sigma_{\varepsilon(D)}^2 > \sigma_{\varepsilon(V)}^2. \\ \text{Measurements are more precise at validation, i.e. } \sigma_{\varepsilon(D)}^2 > \sigma_{\varepsilon(V)}^2. \\ \text{Measurements are more precise at validation, i.e. } \sigma_{\varepsilon(D)}^2 > \sigma_{\varepsilon(V)}^2. \\ \text{Measurements are more precise at validation, i.e. } \sigma_{\varepsilon(D)}^2 > \sigma_{\varepsilon(V)}^2. \\ \text{Measurements are more precise at vali$