# simulation summary binary

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## Simulation study

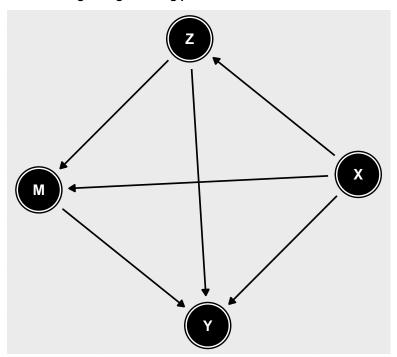
#### Simulation scenarios

Simulations of bias were done with model misspecification and confounding in separate comparable scenarios. 10000 iterations (1/10th of the linear case due to computation time) with a sample size of 2000 were used for each scenario.

Table 1: Variables used for simulations

variable	type	true model
X(additional covariate) Z(exposure)	continuous binary	$X \sim gamma(8, 4.5)$ $Z = I(Z*>0)$ where $Z*\sim U_0 + U_1X + N(0, 1)$
M(mediator)	binary	$M = I(M*>0)$ where $M*\sim \beta_0 + \beta_1 Z + \beta_2 X + N(0,1)$
Y(outcome)	binary	Y = I(Y * > 0)  where $Y = \alpha_0 + \alpha_1 Z + \alpha_2 M + \alpha_3 ZM + \alpha_4 X + N(0, 1)$

#### DAG showing data generating process



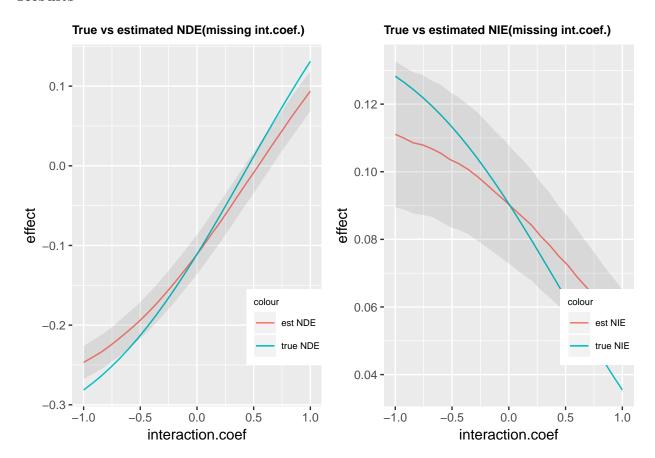
#### Simulation scenarios (model misspecification)

To get realistic parameter values, models were estimated using data from riksstroke from n=60353 patients. Parameter values used for simulations:

```
covariate.parameters = list(c(104, 8, 4.5)), true.exposure.coefs = c(I = -3.416096, X = 0.036231), true.mediator.coefs = c(I = 1, Z = -0.4, X = -0.008, ZX = 0), true.outcome.coefs = c(I = 4, Z = -0.4, M = -2, ZM = corr.coef[i], X = -0.03, ZX = 0, MX = 0, ZMX = 0),
```

Estimated mediator model was set to the correct one. Estimated outcome model was misspecified without ZM interaction:  $Y*\sim Z+M+X$ .

### Results



Total effect is biased. This differs from the situation where we used linear models:



