Simulation Result for Partially Linear Model

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1 Simulation Design

Let

$$y_{i} = d_{i} + x'_{i}(c_{y}\theta_{0}) + u_{i},$$

$$d_{i} = \frac{exp\{x'_{i}(c_{d}\theta_{0})\}}{1 + exp\{x'_{i}(c_{d}\theta_{0})\}} + v_{i},$$

where $x_i, v_i \sim N(0, 1)$, u_i and v_i are independent, $p = dim(x_i) = 250$, the covariates $x_i \sim N(0, \Sigma)$ with $\Sigma_{kj} = (0.5)^{|j-k|}$, and sample size n = 200. θ_0 is a $p \times 1$ vector with elements set as $\theta_{0,j} = (1/j)^2$ for j = 1, ..., p. c_d and c_y are scalars that control the strength of the relationship between the controls, the outcome, and the treatment variables d_i . We can try several different combinations of c_d and c_y , setting

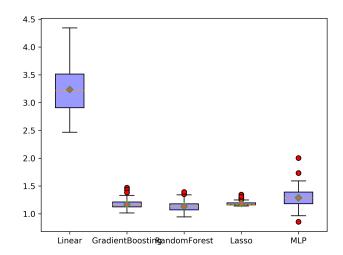
$$c_d = \sqrt{\frac{(\pi^2/3)R_d^2}{(1 - R_d^2)\theta_0'\Sigma\theta_0}}, c_y = \sqrt{\frac{R_y^2}{(1 - R_y^2)\theta_0'\Sigma\theta_0}},$$

for different combinations of $R_d^2, R_y^2 \in \{0, 0.1, 0.5, 0.9\}.$

2 Results

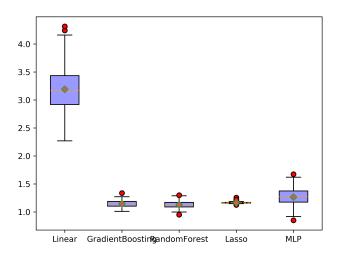
We set $R_d^2 = R_y^2 = 0.1$. 2 folds:

Method	DML1/DML2	$\check{ heta}$	95%CI	$\hat{ heta}$	true θ
LinearRegression	DML1	3.19176	[3.16102, 3.22249]	104.27364	1
LinearRegression	DML2	3.17220	[3.12759, 3.21671]	102.29749	1
${\bf Gradient Boosting Regressor}$	DML1	1.17759	[1.15991, 1.19528]	80.22819	1
${\bf Gradient Boosting Regressor}$	DML2	1.10664	[1.09871, 1.11458]	78.15933	1
RandomForestRegressor	DML1	1.17748	[1.12351, 1.15795]	80.28174	1
RandomForestRegressor	DML2	1.19961	[1.18952, 1.20970]	80.13725	1
Lasso	DML1	1.17748	[1.15791, 1.19705]	88.37562	1
Lasso	DML2	1.154701	[1.14808,1.16134]	87.97575	1
MLPRegressor	DML1	1.29603	[1.28245,1.30962]	88.55359	1
MLPRegressor	DML2	1.25332	[1.24761,1.25904]	86.45241	1



K=2.png

Figure 1: K=2,DML1

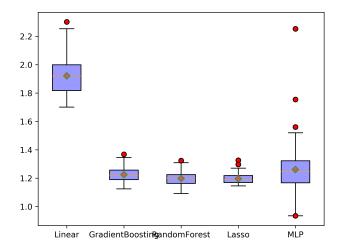


K=2.png

Figure 2: K=2,DML2

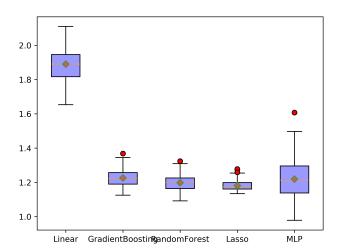
5 folds:

Method	DML1/DML2	$\check{ heta}$	95%CI	$\hat{ heta}$	true θ
LinearRegression	DML1	1.91196	[1.86687, 1.95705]	150.23473	1
LinearRegression	DML2	1.89027	[1.87604, 1.90449]	148.84999	1
${\bf Gradient Boosting Regressor}$	DML1	1.31641	[1.28118, 1.35163]	145.15192	1
${\bf Gradient Boosting Regressor}$	DML2	1.20509	[1.19359, 1.21659]	145.85326	1
RandomForestRegressor	DML1	1.26268	[1.22493,1.30043]	145.41695	1
RandomForestRegressor	DML2	1.21355	[1.20242, 1.22467]	145.00807	1
Lasso	DML1	1.29486	[1.25086, 1.33887]	144.14305	1
Lasso	DML2	1.17356	[1.15417, 1.19295]	140.20278	1
MLPRegressor	DML1	1.32824	[1.26351, 1.39297]	133.92810	1
MLPRegressor	DML2	1.22952	[1.17064, 1.28841]	132.74614	1



K=5.png

Figure 3: K=5,DML1



K=5.png

Figure 4: K=5,DML2