High-dimensional Mixed Linear Regression

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June 26, 2018

Overview

Problem Setting

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Problem Setting : High-dimensional Mixed Linear Regression

- $y_i = \langle x_i, \beta_1^{\star} \rangle z_i + \langle x_i, \beta_2^{\star} \rangle (1 z_i) + w_i$ for $i = 1, \dots, n$
- $x_i, \beta_i^{\star} \in \mathbb{R}^d, z_i \in \{0, 1\}, w_i \stackrel{iid}{\sim} \mathcal{N}(0, \sigma^2)$
- $d \gg n, \beta_k^{\star}$ is sparse
- **Goal**: infer $\beta_1^{\star}, \beta_2^{\star}$ given $\{(x_i, y_i)\}$

Simulation Results

- $y_i = \langle x_i, \beta_1^{\star} \rangle z_i + \langle x_i, \beta_2^{\star} \rangle (1 z_i) + w_i$ for $i = 1, \dots, n$
- $x_i \in \mathcal{N}(0, I_d), \beta_k^{\star} \in \mathbb{R}^d, \|\beta_k^{\star}\|_0 = s, w_i \stackrel{iid}{\sim} \mathcal{N}(0, \sigma^2)$
- $z_i \in \{0,1\}, \sum_i z_i = \lfloor np_1 \rfloor$ (p_1 is a fraction of MLR)
- $r = \min_{k} \{ \| supp(\beta_k^{\star}) \setminus supp(\hat{\beta}) \|_0 \}$ (r is the failure of target recovery)

(n,d,s,p1)	r	(n,d,s,p1)	r	(n,d,s,p1)	r
(3e2,1e4,10,0.5)	4	(6e2,1e4,20,0.5)	8	(9e2,1e4,30,0.5)	20
(4e2,1e4,10,0.5)	3	(8e2,1e4,20,0.5)	3	(1.2e3,1e4,30,0.5)	0
(5e2,1e4,10,0.5)	0	(1e3,1e4,20,0.5)	0	(1.5e3,1e4,30,0.5)	0

Table: $\sigma^2 = 0.1$

Simulation Results

- $y_i = \langle x_i, \beta_1^{\star} \rangle z_i + \langle x_i, \beta_2^{\star} \rangle (1 z_i) + w_i$ for $i = 1, \dots, n$
- $x_i \in \mathcal{N}(0, I_d), \beta_k^{\star} \in \mathbb{R}^d, \|\beta_k^{\star}\|_0 = s, w_i \stackrel{iid}{\sim} \mathcal{N}(0, \sigma^2)$
- $z_i \in \{0,1\}, \sum_i z_i = \lfloor np_1 \rfloor$ (p_1 is a fraction of MLR)
- $r = \min_{k} \{ \| supp(\beta_k^{\star}) \setminus supp(\hat{\beta}) \|_0 \}$ (r is the failure of target recovery)

(n,d,s,σ^2)	r	(n,d,s,σ^2)	r	(n,d,s,σ^2)	r
(5e2,1e4,10,0.1)	0	(1e3,1e4,20,0.1)	0	(1.5e3,1e4,30,0.1)	0
(5e2,1e4,10,0.12)	1	(1e3,1e4,20,0.12)	1	(1.5e3,1e4,30,0.12)	2
(5e2,1e4,10,0.15)	0	(1e3,1e4,20,0.15)	2	(1.5e3,1e4,30,0.15)	3
(5e2,1e4,10,0.2)	1	(1e3,1e4,20,0.2)	0	(1.5e3,1e4,30,0.2)	6
(5e2,1e4,10,0.3)	5	(1e3,1e4,20,0.3)	4	(1.5e3,1e4,30,0.3)	7

Table: $p_1 = 0.5$

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