Logistic Regression with Rmosek

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Consider the simplest logistic regression.

$$\max_{\beta} \sum_{i=1}^{n} \{ y_i (x_i'\beta) - \log (1 + \exp (x_i'\beta)) \}$$
 (1)

Introduce t_i to replace the softplus function in the objective and $\phi_i = x_i'\beta$, we turned (1) to

$$\max_{t_i, \phi_i, \beta} \sum_{i=1}^n \{ y_i \phi_i + t_i \}$$
s.t. $-\log (1 + \exp (\phi_i)) \ge t_i$

$$\phi_i = x_i' \beta$$

Further deal with the softplus function in the constraint by noting that

$$-\log\left(1 + \exp\left(\phi_i\right)\right) \ge t_i \iff \exp\left(\phi_i + t_i\right) + \exp\left(t_i\right) \le 1 \tag{2}$$

Introduce u_i and v_i , (2) can be further transformed to

$$u_i + v_i \le 1$$
$$(u_i, 1, \phi_i + t_i) \in \mathcal{K}_{\text{exp}}$$
$$(v_i, 1, t_i) \in \mathcal{K}_{\text{exp}}$$

where $\mathcal{K}_{\text{exp}} = \{(x_1, x_2, x_3) : x_1 \ge x_2 \exp(x_3/x_2), x_2 > 0\} \cup \{(x_1, 0, x_3) : x_1 \ge 0, x_3 \le 0\}$. We reach the standard format

s.t.
$$\begin{aligned} \max_{\theta} & \begin{bmatrix} \mathbf{1}_n & y & \mathbf{0}_{1 \times (2n+p)} \end{bmatrix} \theta \\ (u_i, 1, \phi_i + t_i) & \in \mathcal{K}_{\text{exp}} i = 1, 2, \cdots, n \end{aligned}$$

where $\theta = (t, \phi, u, v, \beta)$. However it seems that Rmosek accepts only the exact variables in the arguments of the exponential cones, which means we cannot have 1 and $\phi_i + t_i$ in the conic constraints as above. We need

to further introduce $\alpha = \mathbf{1}_{2n}$ and $\gamma = \phi + t$ to reach the final formulation with $\theta = (t, \phi, u, v, \beta, \alpha, \gamma)$.

$$\max_{\theta} \begin{bmatrix} \mathbf{1}_n & y & 0_{1\times(2n+p)} \end{bmatrix} \theta$$
s.t.
$$\begin{bmatrix} 0_{n\times n} & I_n & 0_{n\times 2n} & -X & 0_{n\times(3n)} \\ 0_{n\times 2n} & I_n & I_n & 0_{n\times(p+3n)} \\ I_n & I_n & 0_{n\times(4n+p)} & -I_n \end{bmatrix} = \begin{bmatrix} 0_{n\times 1} \\ \mathbf{1}_n \\ 0_{n\times 1} \end{bmatrix}$$

$$(u_i, \alpha_i, \gamma_i) \in \mathcal{K}_{\exp} \ i = 1, 2, \cdots, n$$

$$(v_j, \alpha_{n+j}, t_j) \in \mathcal{K}_{\exp} \ j = 1, 2, \cdots, n$$

$$\alpha = \mathbf{1}_{2n}$$