

Modelli MPHero

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1 SVR-style models

Data:

- a feature space J
- a set I of points $x^i = (x_1^i \dots x_{|J|}^i)$, and corresponding response values y^i
- a violation cost C
- a confidence region parameter ϵ
- a target number of features to select k_0

Variables:

- hyperplane slopes $w = (w_1 \dots w_{|J|})$ and intercept z values
- feature binary selection decisions f_j (1 if feature j is selected, 0 otherwise)

Optimize:

$$\min \frac{1}{2} \sum_{j \in J} w_j^2 + C \cdot \sum_{i \in I} p_i^+ + p_i^- \quad (1)$$

$$\text{s.t. } \left(\sum_{j \in J} w_j \cdot x_j^i \right) + z - y^i \leq \epsilon + p_i^+ \quad \forall i \in I \quad (2)$$

$$- \left(\sum_{j \in J} w_j \cdot x_j^i \right) - z + y^i \leq \epsilon + p_i^- \quad \forall i \in I \quad (3)$$

$$w_j \leq f_j W_j^U \quad \forall j \in J \quad (4)$$

$$w_j \geq f_j W_j^L \quad \forall j \in J \quad (5)$$

$$\sum_{j \in J} f_j \leq k_0 \quad (6)$$

$$p_i^+ \geq 0, p_i^- \geq 0 \quad \forall i \in I \quad (7)$$

$$f_j \in \{0, 1\} \quad \forall j \in J \quad (8)$$

Where p_i^+, p_i^- are measurement errors on point i .

1.1 Outliers

Optimize:

$$\min \frac{1}{2} \sum_{j \in J} w_j^2 + C \cdot \sum_{i \in I} (p_i^+ + p_i^-) s_i \quad (9)$$

$$\text{s.t. } \left(\sum_{j \in J} w_j \cdot x_j^i \right) + z - y^i \leq \epsilon + p_i^+ \quad \forall i \in I \quad (10)$$

$$- \left(\sum_{j \in J} w_j \cdot x_j^i \right) - z + y^i \leq \epsilon + p_i^- \quad \forall i \in I \quad (11)$$

$$w_j \leq f_j W_j^U \quad \forall j \in J \quad (12)$$

$$w_j \geq f_j W_j^L \quad \forall j \in J \quad (13)$$

$$\sum_{j \in J} f_j \leq k_0 \quad (14)$$

$$\sum_{i \in I} s_i \geq s_0 \quad (15)$$

$$p_i^+ \geq 0, p_i^- \geq 0 \quad \forall i \in I \quad (16)$$

$$f_j \in \{0, 1\} \quad \forall j \in J \quad (17)$$

$$s_i \in \{0, 1\} \quad \forall i \in I \quad (18)$$

The terms $(p_i^+ + p_i^-)s_i$ need to be linearized in the fashion of our previous regression models.

McCormick's linearization. We first perform a standard McCormick's linearization. Note that the linearization of products $p_i^+ s_i$ and $p_i^- s_i$ can be performed in the same way. Then, for the sake of brevity, we denote by $p_i^\pm s_i$ either product. Given a valid upper-bound R_i^\pm on p_i^\pm , and since $s_i \in \{0, 1\}$, it is well-known [?] that for every i we can replace $p_i^\pm s_i$ by an additional variable $t_i \geq 0$ and adding the following 3 sets of constraints for every i :

$$\begin{aligned} t_i^\pm &\leq p_i^\pm \\ t_i^\pm &\leq s_i R_i^\pm \\ t_i^\pm &\geq p_i^\pm - (1 - s_i) R_i^\pm \end{aligned}$$

we obtain a MILP whose solutions corresponds to those of (9)–(18). Hence a first linearization of (9)–(18) is:

$$\min \frac{1}{2} \sum_{j \in J} w_j^2 + C \cdot \sum_{i \in I} (t_i^+ + t_i^-) \quad (19)$$

$$\text{s.t. } \left(\sum_{j \in J} w_j \cdot x_j^i \right) + z - y^i \leq \epsilon + p_i^+ \quad \forall i \in I \quad (20)$$

$$- \left(\sum_{j \in J} w_j \cdot x_j^i \right) - z + y^i \leq \epsilon + p_i^- \quad \forall i \in I \quad (21)$$

$$t_i^+ \leq p_i^+ \quad \forall i \in I \quad (22)$$

$$t_i^+ \leq s_i R_i^+ \quad \forall i \in I \quad (23)$$

$$t_i^+ \geq p_i^+ - (1 - s_i) R_i^+ \quad \forall i \in I \quad (24)$$

$$t_i^- \leq p_i^- \quad \forall i \in I \quad (25)$$

$$t_i^- \leq s_i R_i^- \quad \forall i \in I \quad (26)$$

$$t_i^- \geq p_i^- - (1 - s_i) R_i^- \quad \forall i \in I \quad (27)$$

$$w_j \leq f_j W_j^U \quad \forall j \in J \quad (28)$$

$$w_j \geq f_j W_j^L \quad \forall j \in J \quad (29)$$

$$\sum_{j \in J} f_j \leq k_0 \quad (30)$$

$$\sum_{i \in I} s_i \geq s_0 \quad (31)$$

$$p_i^+ \geq 0, p_i^- \geq 0 \quad \forall i \in I \quad (32)$$

$$f_j \in \{0, 1\} \quad \forall j \in J \quad (33)$$

$$s_i \in \{0, 1\} \quad \forall i \in I \quad (34)$$

$$t_i^+, t_i^- \geq 0 \quad \forall i \in I \quad (35)$$

Disjunctive linearization (Projected McCormick's linearization). We observe that there exists an optimal solution to (19)–(35) satisfying (24) with equality for every $i \in I$. Indeed, if $s_i = 0$ then, given the validity of R_i^\pm , we may always set $p_i^\pm = R_i^\pm$ in an optimal solution (whose value depends only on the t_i^\pm values); when $s_i = 1$ constraint (24) boils down to $t_i^\pm \geq p_i^\pm$, which together with (25), gives $t_i^\pm = p_i^\pm$. So, we can restrict the McCormick's linearization to the set of solutions satisfying (24) with equality. This allows a projection of the t_i^\pm 's variables. After substitution we obtain the following valid MILP for the

starting problem:

$$\min \frac{1}{2} \sum_{j \in J} w_j^2 + C \cdot \sum_{i \in I} (p_i^+ - (1 - s_i)R_i^+) + C \cdot \sum_{i \in I} (p_i^- - (1 - s_i)R_i^-) \quad (36)$$

$$\text{s.t. } \left(\sum_{j \in J} w_j \cdot x_j^i \right) + z - y^i \leq \epsilon + p_i^+ \quad \forall i \in I \quad (37)$$

$$- \left(\sum_{j \in J} w_j \cdot x_j^i \right) - z + y^i \leq \epsilon + p_i^- \quad \forall i \in I \quad (38)$$

$$R_i^+(1 - s_i) \leq p_i^+ \leq R_i^+ \quad \forall i \in I \quad (39)$$

$$R_i^-(1 - s_i) \leq p_i^- \leq R_i^- \quad \forall i \in I \quad (40)$$

$$w_j \leq f_j W_j^U \quad \forall j \in J \quad (41)$$

$$w_j \geq f_j W_j^L \quad \forall j \in J \quad (42)$$

$$\sum_{j \in J} f_j \leq k_0 \quad (43)$$

$$\sum_{i \in I} s_i \geq s_0 \quad (44)$$

$$p_i^+ \geq 0, p_i^- \geq 0 \quad \forall i \in I \quad (45)$$

$$f_j \in \{0, 1\} \quad \forall j \in J \quad (46)$$

$$s_i \in \{0, 1\} \quad \forall i \in I \quad (47)$$

1.2 The Lagrangian Dual of the Disjunctive Linearization Model

In order to easily describe the dual formulation we introduce and fix some notations:

- v represents a generic variables (that is, any of $w_j, z, p_i^\pm, f_j, s_i$) and \mathbf{v} is their vector (following the ordering $w_j, z, p_i^\pm, f_j, s_i$).
- $\Gamma \mathbf{v} \leq \gamma$ is the constraint matrix of the available generic cuts obtained from the ILP. The column corresponding to variable v is indicated with Γ^v . Then Γ_h^v is the h -th entry of that column. We assume that the matrix Γ has $m \geq 0$ row.

- The x^i 's are column vectors in \mathbb{R}^d , so that x_j^i is the j -th entry of x^i .

We define the duals of the disjunctive-based formulation.

- α^+ (resp. α^-) is the dual vector of constraints (37) (resp. (38))
- π^+ and ψ^+ (resp. π^- and ψ^-) are the dual vectors of upper- and lower-bound constraints (39) (resp. (40))
- λ^U and λ^L are the dual vectors of constraints (41) and (42) respectively
- β_1 and β_2 are the duals of constraints (43) and (44)
- η^+ and η^- are the dual vectors of constraints (45)
- φ (resp. σ) are the dual vectors of upper- and lower bound constraints $f_j \leq 1$ and $f_j \geq 0$ (resp. $s_i \leq 1$ and $s_i \geq 0$).

We have that the dual of the disjunctive-based formulation is:

$$\begin{aligned}
\min \quad & \frac{1}{2} \sum_{j \in J} \left(\sum_{i \in I} x_j^i (\alpha_i^- - \alpha_i^+) \right)^2 + \sum_{j \in J} \left(\left(\lambda_j^L - \lambda_j^U - \sum_{h=1}^m \xi_h \Gamma_h^{\omega_j} \right) \sum_{i \in I} x_j^i (\alpha_i^- - \alpha_i^+) \right) + \frac{1}{2} \sum_{j \in J} \left(\lambda_j^L - \lambda_j^U - \sum_{h=1}^m \xi_h \Gamma_h^{\omega_j} \right)^2 \\
& - \sum_{i \in I} (\alpha_i^+ (y_i + \epsilon) - \alpha_i^- (y_i - \epsilon)) - \sum_{i \in I} ((\pi_i^+ - \psi_i^+) R_i^+ + (\pi_i^- - \psi_i^-) R_i^-) - k_0 \beta_1 + s_0 \beta_2 \\
& - \sum_{j \in J} \varphi_j - \sum_{i \in I} \sigma_i - \sum_{h=1}^m \xi_h \gamma_h \tag{48}
\end{aligned}$$

s.t.

$$\sum_{i \in I} (\alpha_i^+ - \alpha_i^-) + \sum_{h=1}^m \xi_h \Gamma_h^z = 0 \tag{dz}$$

$$\psi_i^+ + \alpha_i^+ - \pi_i^+ - \sum_{h=1}^m \xi_h \Gamma_h^{p_i^+} \leq C \quad \forall i \in I \tag{dp_i^+}$$

$$\psi_i^- + \alpha_i^- - \pi_i^- - \sum_{h=1}^m \xi_h \Gamma_h^{p_i^-} \leq C \quad \forall i \in I \tag{dp_i^-}$$

$$\lambda_j^L W_j^L - \lambda_j^U W_j^U + \varphi_j + \beta_1 + \sum_{h=1}^m \xi_h \Gamma_h^{f_j} \geq 0 \quad \forall j \in J \tag{df_j}$$

$$\psi_i^+ R_i^+ + \psi_i^- R_i^- + \beta_2 - \sigma_i - \sum_{h=1}^m \xi_h \Gamma_h^{s_i} \leq C (R_i^+ + R_i^-) \quad \forall i \in I \tag{ds_i}$$

$$\alpha^\pm, \pi^\pm, \psi^\pm, \lambda^U, \lambda^L, \beta_1, \beta_2, \varphi, \sigma, \xi \geq \mathbf{0} \tag{49}$$