## L1-Norm Minimization

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**Theorem 1:** Given two vectors  $\mathbf{x}, \mathbf{y} \in \mathbb{R}^n$ , the optimal scalar  $\boldsymbol{\beta}$  that minimizes  $||\boldsymbol{\beta}\mathbf{x} - \mathbf{y}||_1$  is

$$\beta^{\star} = \frac{y_{j_{t^{\star}}}}{x_{j_{t^{\star}}}},$$

where:

- 1.  $x_i$  or  $y_i$  is the *i*-th element of  $\boldsymbol{x}$  or  $\boldsymbol{y}$ , respectively;
- 2.  $\{j_1, \ldots, j_m\}$  is a permutation of the indices of non-zero elements in  $\boldsymbol{x}$  ( $\{i_1, \ldots, i_m\}$ , where  $m \le n$ ) satisfying  $\frac{y_{j_1}}{x_{j_1}} \le \ldots \le \frac{y_{j_m}}{x_{j_m}}$ ;
- 3.  $t^*$  satisfies  $\sum_{k=1}^{t} |x_{j_k}| \sum_{k=t+1}^{m} |x_{j_k}| < 0$  for  $t < t^*$  and  $\sum_{k=1}^{t} |x_{j_k}| \sum_{k=t+1}^{m} |x_{j_k}| \ge 0$  for  $t^* \le t \le m$ .

*Proof:*  $f(\beta) = ||\beta \mathbf{x} - \mathbf{y}||_1$  is equivalent to

$$f(\beta) = \sum_{i=1}^{n} |\beta x_i - y_i| = \sum_{k=1}^{m} |\beta x_{i_k} - y_{i_k}|$$

$$= \sum_{k=1}^{m} |x_{i_k}| \times \left| \beta - \frac{y_{i_k}}{x_{i_k}} \right| = \sum_{k=1}^{m} |x_{j_k}| \times \left| \beta - \frac{y_{j_k}}{x_{j_k}} \right|.$$
(1)

First, we show the optimal value of  $\beta$  is one of  $\frac{y_{j_1}}{x_{j_1}}, \dots, \frac{y_{j_m}}{x_{j_m}}$ .

When  $\beta \leq \frac{y_{j_1}}{x_{j_1}}$ ,  $f(\beta)$  in (1) is equal to

$$f(\beta) = \sum_{k=1}^{m} |x_{j_k}| \times \left(\frac{y_{j_k}}{x_{j_k}} - \beta\right) = \sum_{k=1}^{m} |x_{j_k}| \times \frac{y_{j_k}}{x_{j_k}} - \left(\sum_{k=1}^{m} |x_{j_k}|\right) \times \beta.$$
 (2)

The minimum of (2) is attained when  $\beta = \frac{y_{j_1}}{x_{j_1}}$ . Similarly, when  $\beta \ge \frac{y_{j_m}}{x_{j_m}}$ , we obtain

$$f(\beta) = \sum_{k=1}^{m} \left| x_{j_k} \right| \times \left( \beta - \frac{y_{j_k}}{x_{j_k}} \right) = \left( \sum_{k=1}^{m} \left| x_{j_k} \right| \right) \times \beta - \sum_{k=1}^{m} \left| x_{j_k} \right| \times \frac{y_{j_k}}{x_{j_k}}. \tag{3}$$

The minimum of (3) is achieved when  $\beta = \frac{y_{jm}}{x_{jm}}$ .

When  $\frac{y_{j_t}}{x_{j_t}} \le \beta \le \frac{y_{j_{t+1}}}{x_{j_{t+1}}}$  for  $1 \le t \le m-1$ , we have

$$f(\beta) = \sum_{k=1}^{t} |x_{j_k}| \times \left(\beta - \frac{y_{j_k}}{x_{j_k}}\right) + \sum_{k=t+1}^{m} |x_{j_k}| \times \left(\frac{y_{j_k}}{x_{j_k}} - \beta\right)$$

$$= \left(\sum_{k=1}^{t} |x_{j_k}| - \sum_{k=t+1}^{m} |x_{j_k}|\right) \times \beta + \sum_{k=t+1}^{m} |x_{j_k}| \times \frac{y_{j_k}}{x_{j_k}} - \sum_{k=1}^{t} |x_{j_k}| \times \frac{y_{j_k}}{x_{j_k}}.$$
(4)

The minimum of (4) is obtained when  $\beta = \frac{y_{j_t}}{x_{j_t}}$  if  $\sum_{k=1}^t \left| x_{j_k} \right| - \sum_{k=t+1}^m \left| x_{j_k} \right| \ge 0$  and  $\beta = \frac{y_{j_{t+1}}}{x_{j_{t+1}}}$  if  $\sum_{k=1}^t \left| x_{j_k} \right| - \sum_{k=t+1}^m \left| x_{j_k} \right| < 0$ . Therefore, we have shown the optimal  $\beta$  should be one of  $\frac{y_{j_1}}{x_{j_1}}, \ldots, \frac{y_{j_m}}{x_{j_m}}$ .

Second, we show the optimal  $\boldsymbol{\beta}$  is  $\boldsymbol{\beta}^{\star} = \frac{y_{j_t \star}}{x_{j_t \star}}$  such that  $\sum_{k=1}^{t} \left| x_{j_k} \right| - \sum_{k=t+1}^{m} \left| x_{j_k} \right| < 0$  for  $t < t^{\star}$  and  $\sum_{k=1}^{t} \left| x_{j_k} \right| - \sum_{k=t+1}^{m} \left| x_{j_k} \right| \ge 0$  for  $t^{\star} \le t \le m$ . By subtracting  $f\left(\frac{y_{j_t}}{x_{j_t}}\right)$  from  $f\left(\frac{y_{j_{t+1}}}{x_{j_{t+1}}}\right)$ , we get

$$f\left(\frac{y_{j_{t+1}}}{x_{j_{t+1}}}\right) - f\left(\frac{y_{j_{t}}}{x_{j_{t}}}\right) = \sum_{k=1}^{t} |x_{j_{k}}| \times \left(\frac{y_{j_{t+1}}}{x_{j_{t+1}}} - \frac{y_{j_{k}}}{x_{j_{k}}}\right) + \sum_{k=t+2}^{m} |x_{j_{k}}| \times \left(\frac{y_{j_{k}}}{x_{j_{k}}} - \frac{y_{j_{t+1}}}{x_{j_{t+1}}}\right)$$
$$- \sum_{k=1}^{t-1} |x_{j_{k}}| \times \left(\frac{y_{j_{t}}}{x_{j_{t}}} - \frac{y_{j_{k}}}{x_{j_{k}}}\right) - \sum_{k=t+1}^{m} |x_{j_{k}}| \times \left(\frac{y_{j_{k}}}{x_{j_{k}}} - \frac{y_{j_{t}}}{x_{j_{t}}}\right)$$
$$= \left(\sum_{k=1}^{t} |x_{j_{k}}| - \sum_{k=t+1}^{m} |x_{j_{k}}|\right) \times \left(\frac{y_{j_{t+1}}}{x_{j_{t+1}}} - \frac{y_{j_{t}}}{x_{j_{t}}}\right).$$

Note that  $\frac{y_{j_{t+1}}}{x_{j_{t+1}}} - \frac{y_{j_t}}{x_{j_t}} \ge 0$ . For any  $t < t^\star$ ,  $\sum_{k=1}^t \left| x_{j_k} \right| - \sum_{k=t+1}^m \left| x_{j_k} \right| < 0$ , thus,  $f\left(\frac{y_{j_t}}{x_{j_t}}\right) \ge f\left(\frac{y_{j_{t+1}}}{x_{j_{t+1}}}\right) \ge \cdots \ge f(\beta^\star)$ . For any  $t > t^\star$ ,  $\sum_{k=1}^t \left| x_{j_k} \right| - \sum_{k=t+1}^m \left| x_{j_k} \right| \ge 0$ , thus,  $f\left(\frac{y_{j_t}}{x_{j_t}}\right) \ge f\left(\frac{y_{j_{t-1}}}{x_{j_{t-1}}}\right) \ge \cdots \ge f(\beta^\star)$ . Furthermore, since  $\sum_{k=1}^t \left| x_{j_k} \right| - \sum_{k=t+1}^m \left| x_{j_k} \right|$  is strictly increasing on t, there exists only one such  $t^\star$ . The proof is complete.