

## 1 Subroutine for SDForest

We are fitting a single tree. At step  $m$ , the tree has  $m$  leaves. We encode this by a matrix  $E \in \mathbb{R}^{n \times m}$ . We write  $e_1, \dots, e_m$  for the columns of  $E$ . The matrix  $E$  has the property that if  $1 \leq l < t \leq m$ , either  $e_l$  and  $e_t$  have disjoint support or the support of  $e_t$  is contained in the support of  $e_l$ . To find the best  $(m+1)$ th split, we consider a large number of candidate split encoded by a new column  $e_{m+1}$ , which has a 1 in the  $i$ th entry if the  $i$ th sample point  $x_i$  lies in the new leaf. We want to find the new column such that  $\|QE_{m+1}\hat{\beta}_{m+1} - QY\|_2^2$  is minimal among the candidate splits, with  $E_{m+1} = (E_m, e_{m+1}) \in \mathbb{R}^{n \times (m+1)}$  and  $\hat{\beta}_{m+1}$  is the least squares estimator of  $QY$  vs.  $QE_{m+1}$ . The goal of this note is to show, how we can efficiently find the best split  $e_{m+1}$  without having to estimate a linear model "from scratch" each time.

By induction, assume that we are given a QR-decomposition of the matrix  $QE_m$ , i.e. there exists a matrices  $U_m \in \mathbb{R}^{n \times m}$  and  $R_m \in \mathbb{R}^{m \times m}$  such that the columns  $u_1, \dots, u_m$  of  $U_m$  are orthonormal and  $R_m$  is an upper triangular matrix and

$$QE_m = U_m R_m.$$

For a candidate split encoded by  $e_{m+1}$ , let  $w_{m+1} = Qe_{m+1}$ . Define

$$u'_{m+1} = w_{m+1} - (w_{m+1}^T u_1)u_1 - \dots - (w_{m+1}^T u_m)u_m.$$

Then, define  $u_{m+1} = u'_{m+1} / \|u'_{m+1}\|$ . Note that  $u_{m+1}$  is orthogonal to  $u_1, \dots, u_m$ . Moreover,  $w_{m+1}$  is in the span of  $u_1, \dots, u_{m+1}$  and  $w_{m+1} = (w_{m+1}^T u_1)u_1 + \dots + (w_{m+1}^T u_{m+1})u_{m+1}$ . Define  $r_{m+1} = (w_{m+1}^T u_1, \dots, w_{m+1}^T u_{m+1})^T \in \mathbb{R}^{m+1}$ . In total, we can write

$$QE_{m+1} = U_{m+1} R_{m+1},$$

where  $U_{m+1}$  has orthonormal columns  $u_1, \dots, u_{m+1}$  and

$$R_{m+1} = \begin{pmatrix} R_m & r_{m+1} \\ 0 & \vdots \end{pmatrix}$$

is an upper triangular matrix. The least squares estimator  $\hat{\beta}_{m+1} = \arg \min_{\beta} \|QE_{m+1}\beta - QY\|^2$  is given by

$$\begin{aligned} \hat{\beta}_{m+1} &= ((QE_{m+1})^T QE_{m+1})^{-1} (QE_{m+1})^T QY \\ &= (R_{m+1}^T U_{m+1}^T U_{m+1} R_{m+1})^{-1} R_{m+1}^T U_{m+1}^T QY \\ &= R_{m+1}^{-1} U_{m+1}^T QY \end{aligned}$$

We are interested in choosing  $e_{m+1}$  such that  $\|QE_{m+1}\hat{\beta}_{m+1} - QY\|_2^2$  is minimal. Note that

$$\begin{aligned}
\|QE_{m+1}\hat{\beta}_{m+1} - QY\|_2^2 &= \|U_{m+1}R_{m+1}R_{m+1}^{-1}U_{m+1}^TQY - QY\|_2^2 \\
&= \|U_{m+1}U_{m+1}^TQY - QY\|_2^2 \\
&= (QY)^T(I - U_{m+1}U_{m+1}^T)^2QY \\
&= (QY)^T(I - U_{m+1}U_{m+1}^T)QY \\
&= \|QY\|^2 - \|U_{m+1}^TQY\|^2 \\
&= \|QY\|^2 - \|U_m^TQY\|^2 - (u_{m+1}^TQY)^2
\end{aligned}$$

Hence, we need to choose  $e_{m+1}$  such that  $(u_{m+1}^TQY)^2$  is maximal.

Hence, the algorithm to find the optimal split  $e_{m+1}$  has the following steps:

For all candidate splits  $s$ , let  $e_{m+1}^s$  be the encoding of this split. For all  $s$ , do

1.  $w_{m+1} = Qe_{m+1}^s$
2.  $u'_{m+1} = w_{m+1} - (w_{m+1}^T u_1)u_1 - \dots - (w_{m+1}^T u_m)u_m$
3. Store  $\alpha_s = (u_{m+1}'^T QY)^2 / \|u'_{m+1}\|^2$

Choose  $s$ , such that  $\alpha_s$  is maximal. Then save  $u_{m+1} = u'_{m+1} / \|u'_{m+1}\|$  with the  $u'_{m+1}$  from the optimal  $s$ .

This can again be made faster (note  $Q^T = Q$ ): For all candidate splits  $s$ , let  $e_{m+1} = e_{m+1}^s$  be the encoding of this split. For all  $s$ , do

1.  $u'_{m+1} = Qe_{m+1} - (e_{m+1}^T Qu_1)u_1 - \dots - (e_{m+1}^T Qu_m)u_m$
2. Store  $\alpha_s = (u_{m+1}'^T QY)^2 / \|u'_{m+1}\|^2$

Choose  $s$ , such that  $\alpha_s$  is maximal. Then save  $u_{m+1} = u'_{m+1} / \|u'_{m+1}\|$  with the  $u'_{m+1}$  from the optimal  $s$ . This should be faster since  $Qu_1, \dots, Qu_m$  only have to be calculated once.

This can again be rewritten. We use  $(e_{m+1}^T Qu_j)u_j = u_j u_j^T Qe_{m+1}$ . Hence, we can replace the first line by

1.  $u'_{m+1} = (Q - \sum_{l=1}^m u_l u_l^T Q) e_{m+1}$

Hence,  $(Q - \sum_{l=1}^m u_l u_l^T Q)$  is always the same and only needs to be updated by subtracting  $u_{m+1} u_{m+1}^T Q$ , once the best split is decided. Hence, what is remained to do for each candidate split is really just the matrix vector product  $(Q - \sum_{l=1}^m u_l u_l^T Q) e_{m+1}$  and the scalar product  $(u_{m+1}'^T QY)^2 / \|u'_{m+1}\|^2$  in step 2. This should be faster than the approach with solving a linear model if the second step can be made sufficiently fast.