

2) MARS model is given by $f(x) = \beta_0 + \sum_{m=1}^M \beta_m h_m(x)$ — (1) $h_m(x) = \begin{cases} (x-t) & \text{if } x > t \\ 0 & \text{otherwise} \end{cases}$

where $h_m(x)$ is a piecewise linear basis functions.

Now consider the classification and regression trees algorithm (that is CART) the formula is:

$$f(x) = \sum_{m=1}^M c_m I\{x \in R_m\} \quad \text{where — (2)}$$

where I is a identity function that returns 1 if x is in subset R_m

Thus we can see that MARS to be a modification of CART algorithm with a better regression setting.

Hence by replacing the piecewise linear basis functions by step identity functions $I(x-t > 0)$ & $I(x-t \leq 0)$ i.e.,

$$f(x) = \sum_{m=0}^M \beta_m I_m(x-t) \quad \text{where } I_m(x-t) = \begin{cases} R_1 & x > t \\ R_2 & x \leq t \end{cases}$$

In other words & MARS ~~is~~ multiplicative model h_m is replaced with interaction ~~model~~ method I_m , a reflected step function pair.

Finally, To get a binary tree representation of, the step function

should be restricted to not to split more than more than once.

By following the two steps, we can modify MARS method to behave like a decision tree.

(b) Since MARS use piece-wise ~~low~~ linear basis functions, they are more powerful regression compared to identity step functions. Hence, MARS can ~~of~~ express ~~to~~ represent better the underlying data distribution. better in comparison to binary ~~to~~ trees. Hence for very high dimensioned inputs which is common ~~are~~ in real world applications, MARS method is a better regressor model.

At the same time, the process of adding ~~the~~ a basis function to regressor model in MARS is a very computationally expensive procedure. It can be shown that for N data points E p predictors and m back fitting algorithm cycles,

(i) trees take $2(pN \log N)$ operations. (worst case $pN \log N + N^2 p$)

(ii) For a M -term ~~model~~ MARS model require $NM^3 + pM^2N$ computations, if M is reasonable fraction of N then it is very expensive.

Clearly MARS methods due to its complexity can ~~are~~ get prohibitively expensive compared to decision trees.