

# Nonparametric mfrd

Wenyu Zhang

December 2018

This document describes the nonparametric frontier approach in function *mfrd\_est*:

```
mfrd_est(y, x1, x2, c1, c2, t.design = NULL, local = 0.15, front.bw = NA,  
m = 10, k = 5, kernel = "triangular", ngrid = 250, margin = 0.03, boot =  
NULL, cluster = NULL, stop.on.error = TRUE)
```

## 1 Optimal bandwidth

Assuming that we have an optimal bandwidth  $b^*$ , then estimation of treatment effects can proceed similar to the univariate case. We fit a weighted linear model using only points within  $L_1$  distance  $b^*$  of the interested frontiers, with weights calculated according to the specified kernel. We denote the linear model by  $f(x_1, x_2; b^*)$ .

## 2 Bandwidth evaluation

We want  $b^*$  to be optimal in estimating the treatment effects at the frontiers. To evaluate how good a bandwidth  $b$  is, we use the mean squared error (MSE) for estimation on a test set:

$$\frac{1}{|S(\delta)|} \sum_{(x_1, x_2) \in S(\delta)} (f(x_1, x_2; b) - y)^2$$

where  $f$  is fitted using the training set, and  $S(\delta)$  is the test set where all points are within  $L_1$  distance  $\delta$  of the interest frontiers. This means that:

$$b^* = \operatorname{argmin}_b \frac{1}{|S(\delta)|} \sum_{(x_1, x_2) \in S(\delta)} (f(x_1, x_2; b) - y)^2$$

Since we have three treatment effect models (i.e. complete, heterogeneous treatments, treatment only), there is a MSE and hence optimal bandwidth corresponding to each.

Since it is difficult to optimize for  $b^*$  exactly, we select the best  $b$  from a random sample. In the *mfrd\_est* function, we draw  $m$  choices of  $b$  uniformly-at-random from the interval  $[0.5, 2.5]$  for the standardized  $x_1$  and  $x_2$ , and  $m = 10$

as the default value. We set  $\delta = 0.25$  to focus on effects at the frontier and also to provide fairer comparison among different  $b$ 's.

### 3 Cross-validation for MSE

To calculate the MSE, we implement  $k$ -fold cross-validation, with  $k = 5$  as the default. In each iteration, the  $k$ -th set is used to produce  $S(\delta)$ , and the remaining  $k - 1$  sets are used to train the linear model  $f$ . The final MSE is the average across all  $k$ -folds, and the optimal empirical bandwidth is chosen as the minimizer of this MSE.