

```
$project
```

```
[1] "DA"
```

```
$platform
```

```
[1] "google_play"
```

```
$dig
```

```
[1] 3
```

```
$dataFile
```

```
[1] "payer_model_DA_GP&iOS_mkt_2019-04-01_2019-06-30.rds"
```

```
$sampleSize
```

```
[1] 8e+05
```

```
$testSampleSplit
```

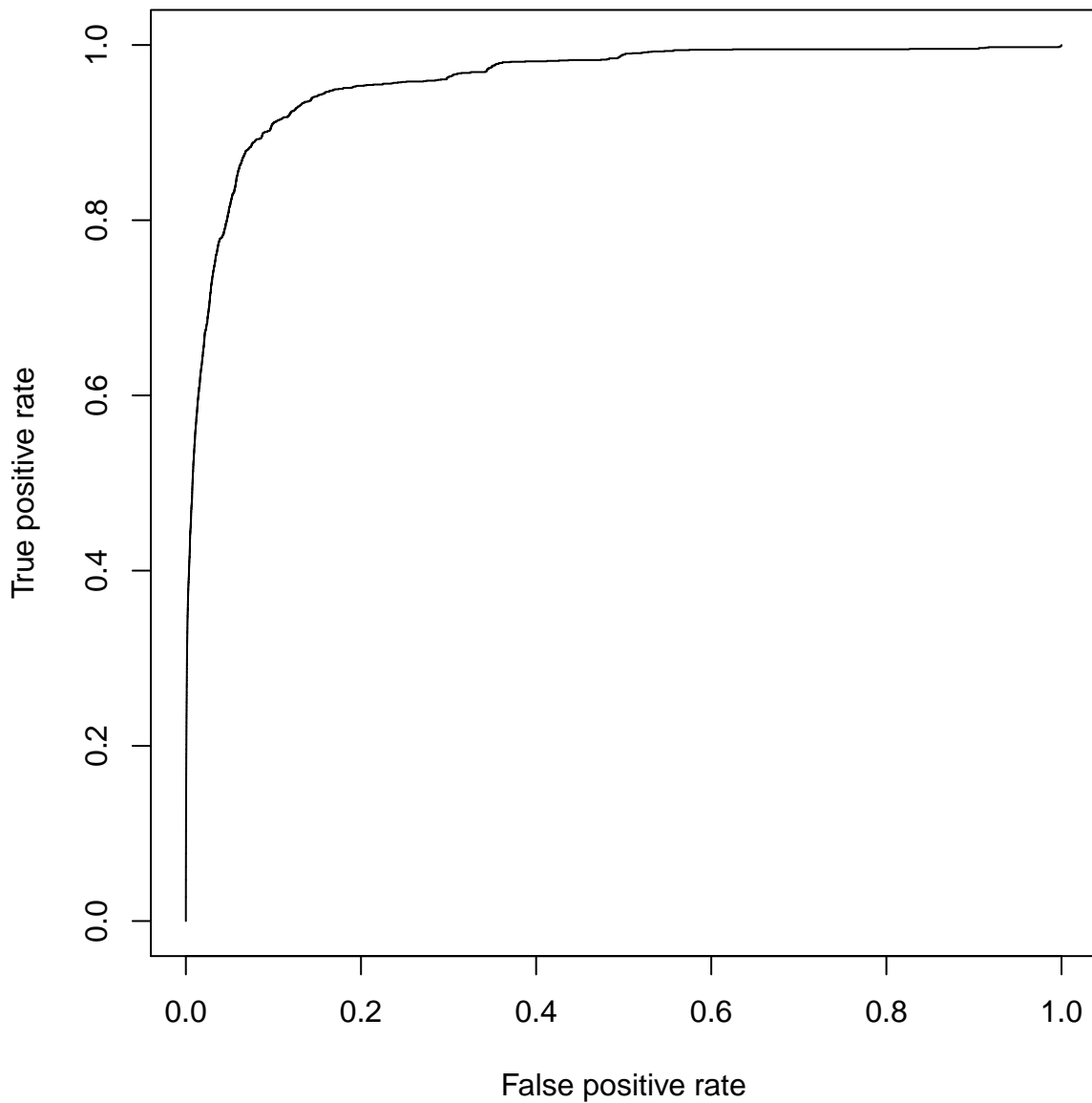
```
[1] 0.5
```

```
$seed
```

```
[1] 1
```

Transformation by Box-Cox together with scaling seems to slightly improve the performance of the model. We should run CV to be sure, since this is only one sample. But transformation definitely has some potential, since we can see in the histogram that the transformed fit better distinguishes between positives and negatives.

**basic model AUC = 0.959**



```
$note  
[1] "no scaling or tranformation"
```

```
$rcd_optimal_cut_off  
[1] 0.1312741
```

```
$rcd_optimal_confusion_matrix
```

	FALSE	TRUE
FALSE	396739	1252
TRUE	1222	787

```
$sensitivity  
[1] 0.3859735
```

```
$precision  
[1] 0.3917372
```

```
$relative_count_difference  
[1] 0.9852869
```

FALSE



TRUE



fit

600

500

400

300

200

100

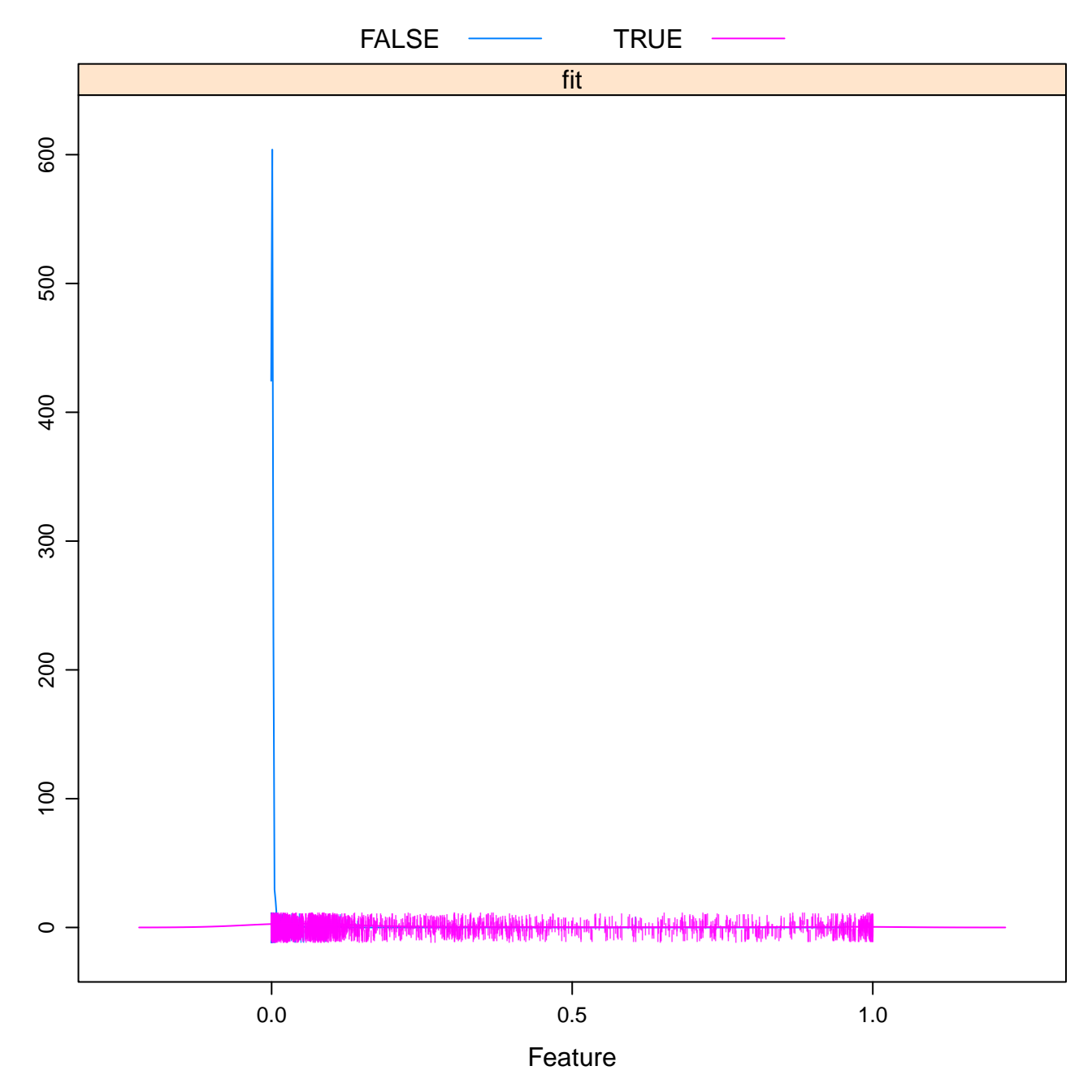
0

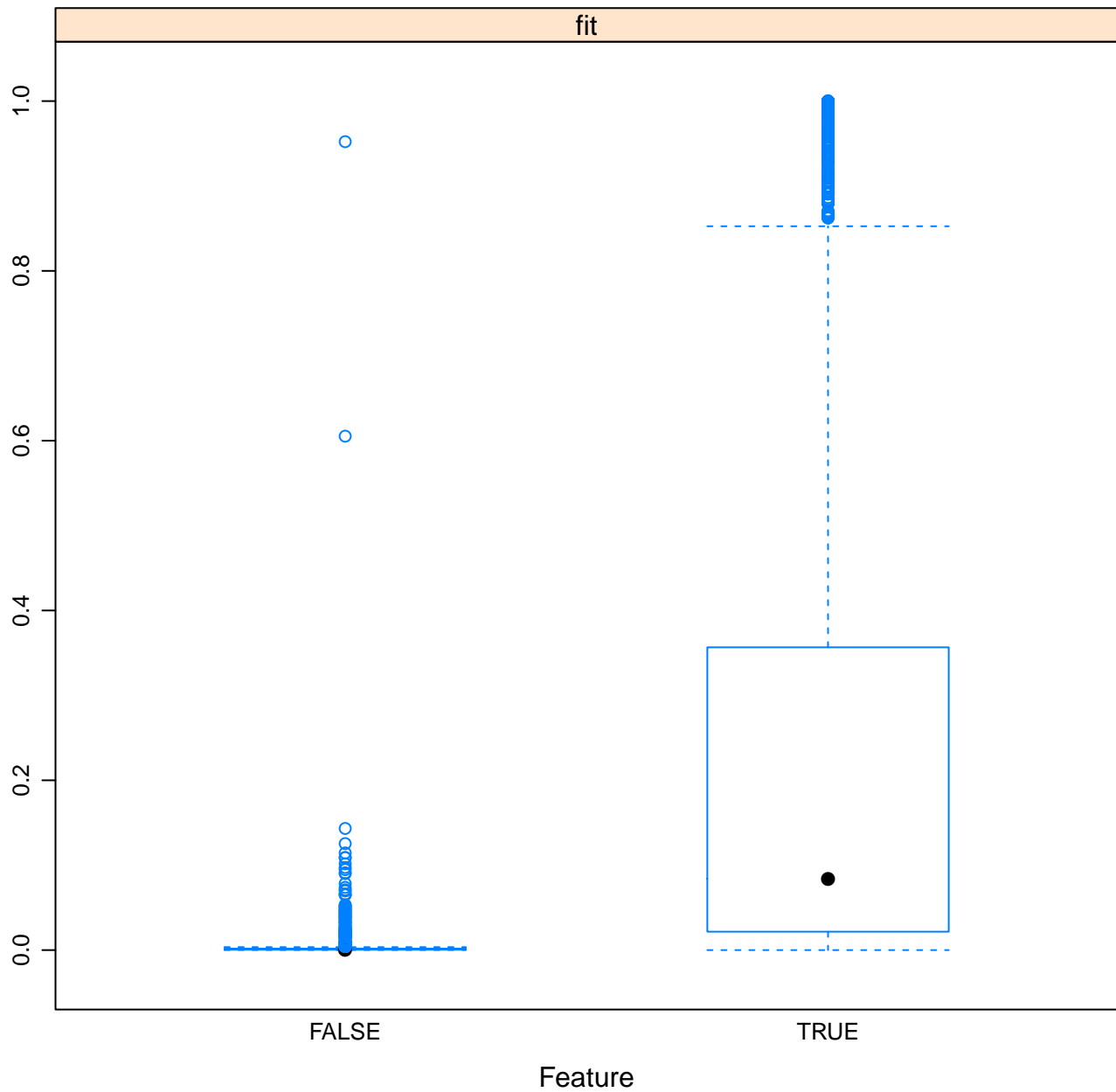
0.0

0.5

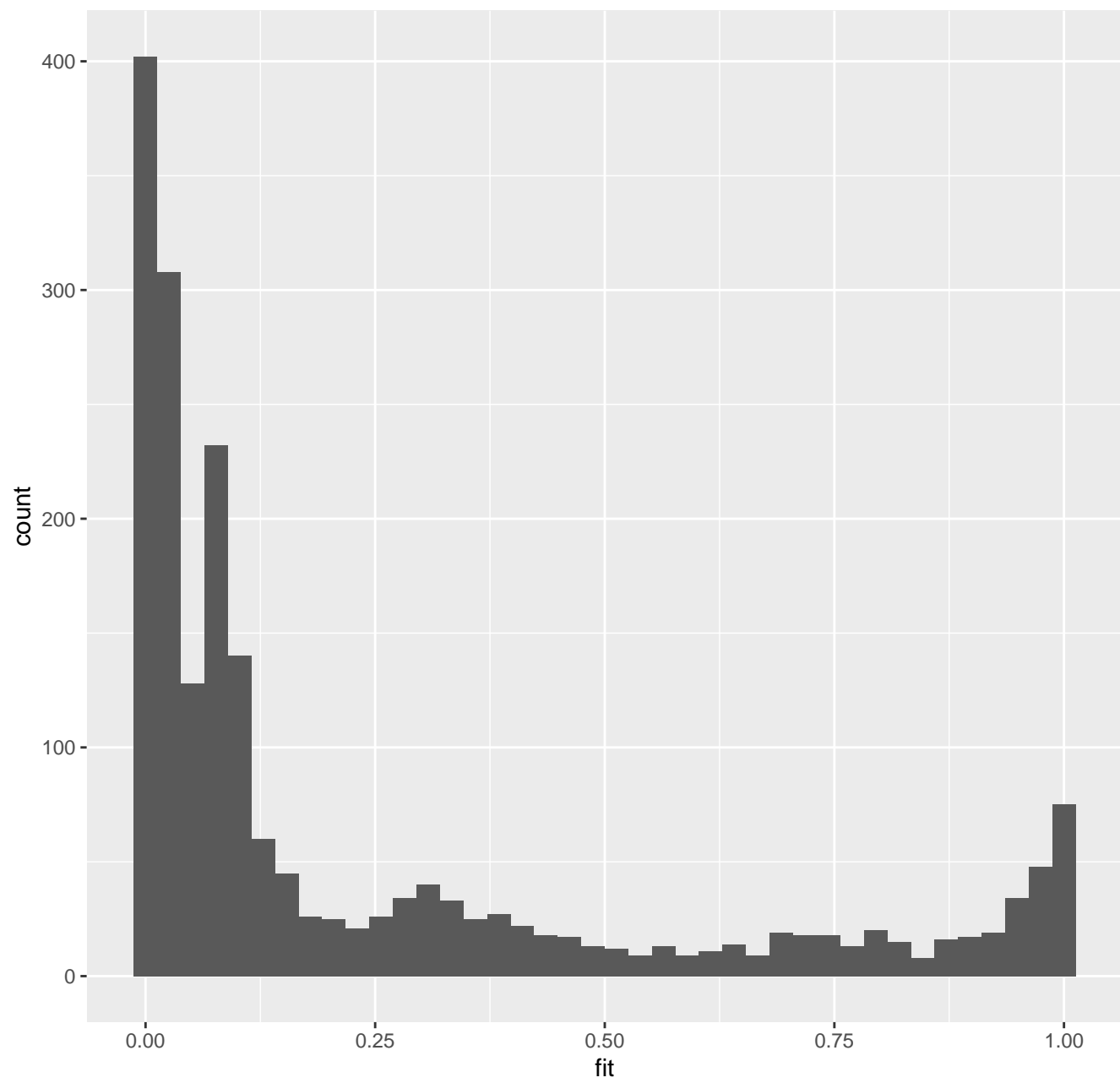
1.0

Feature

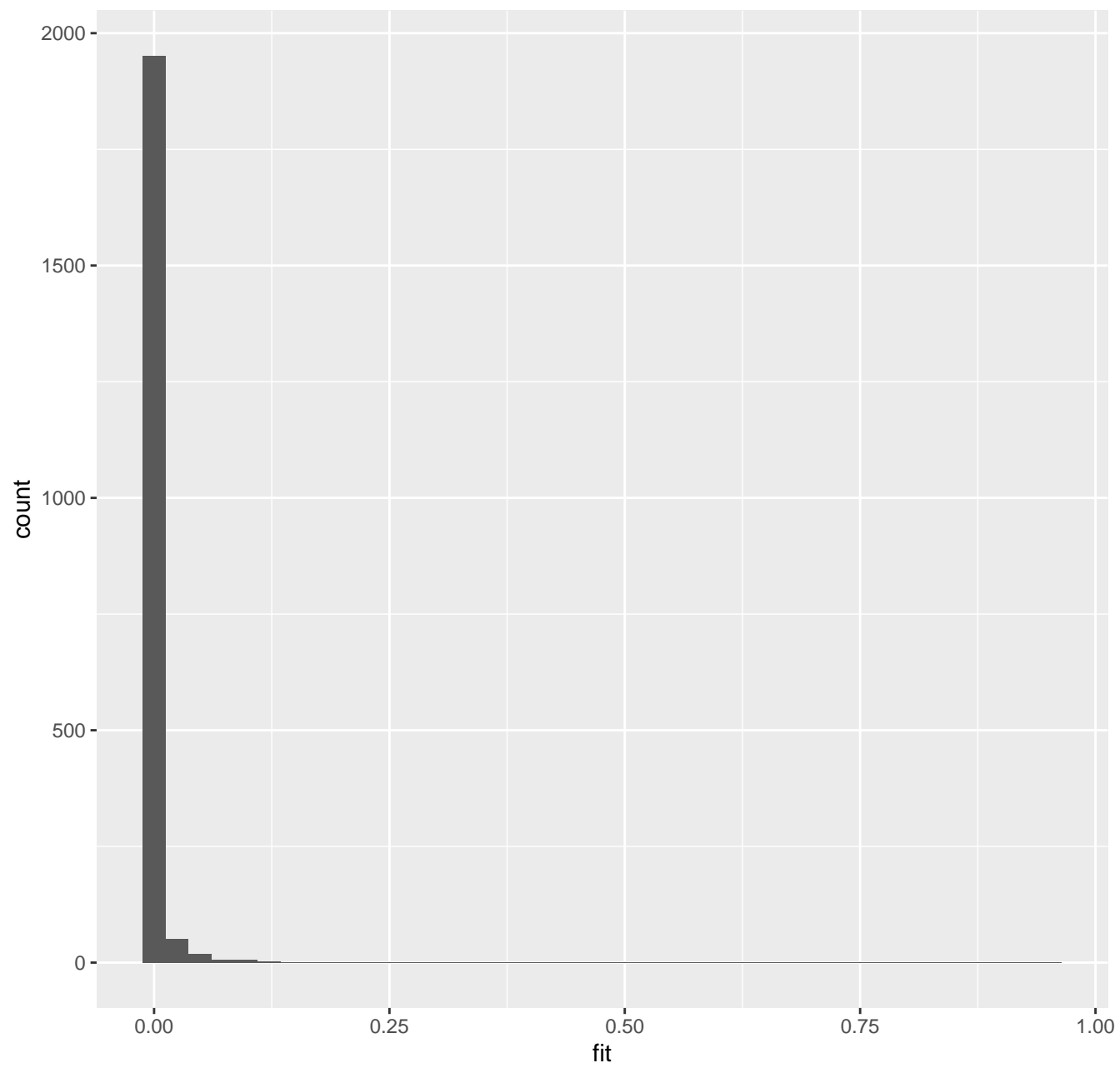




dy\_payer == TRUE

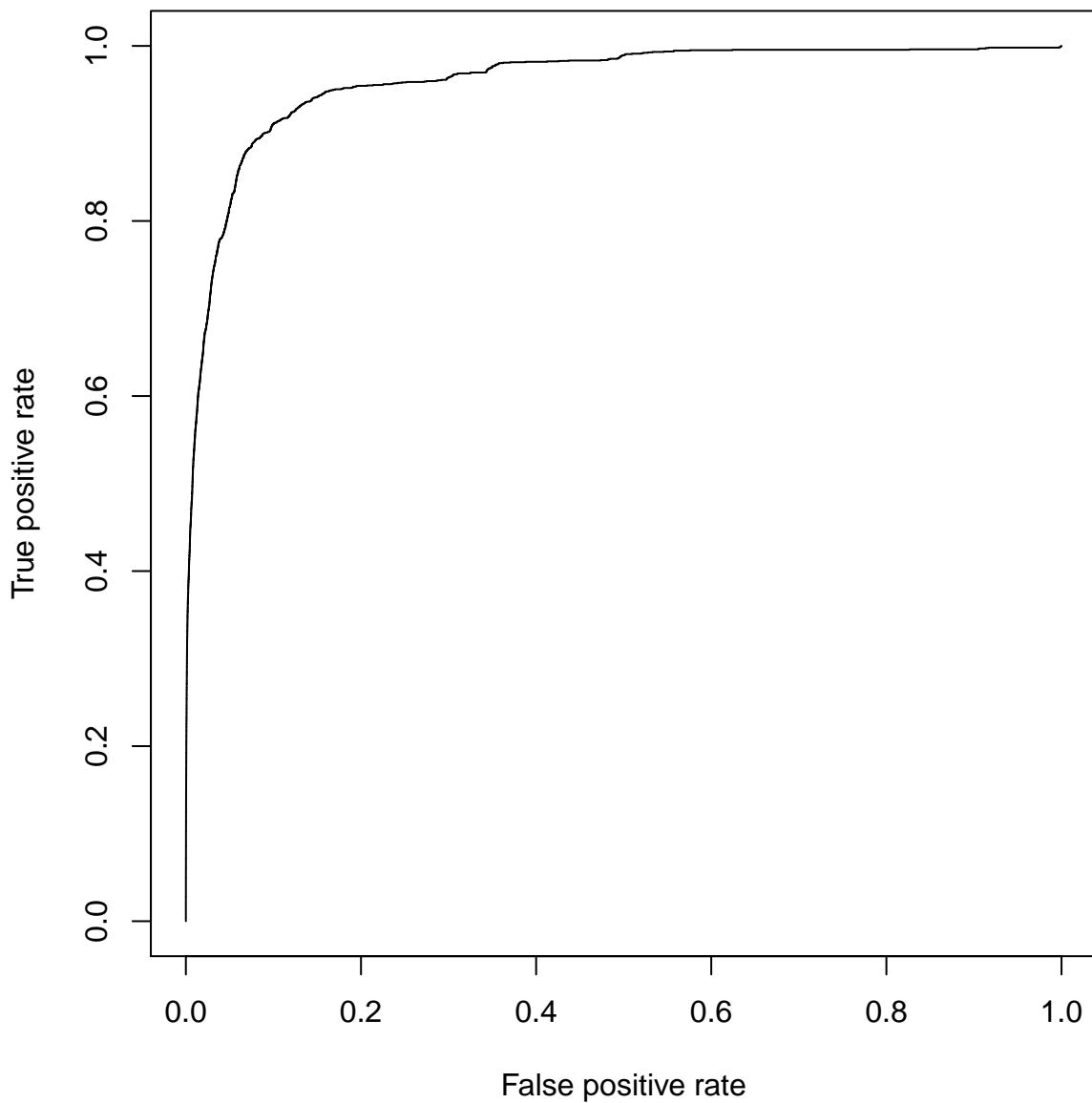


dy\_payer == TRUE





**scaled model AUC = 0.96**



```
$note  
[1] "scaled/standardized features"
```

```
$rcd_optimal_cut_off  
[1] 0.1312741
```

```
$rcd_optimal_confusion_matrix
```

	FALSE	TRUE
FALSE	396655	1235
TRUE	1306	804

```
$sensitivity  
[1] 0.3943109
```

```
$precision  
[1] 0.3810427
```

```
$relative_count_difference  
[1] 1.034821
```

FALSE



TRUE



fit

600

500

400

300

200

100

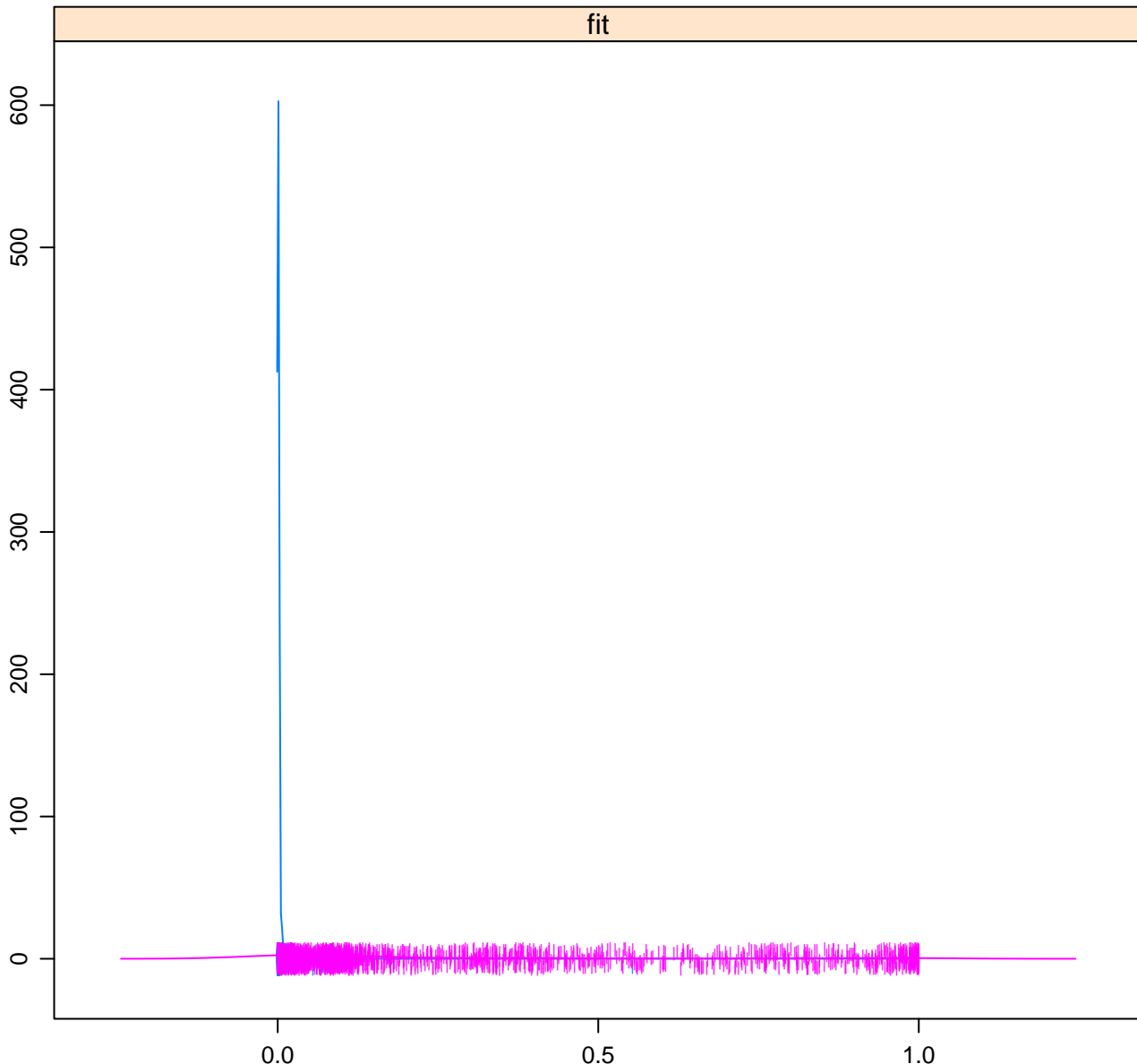
0

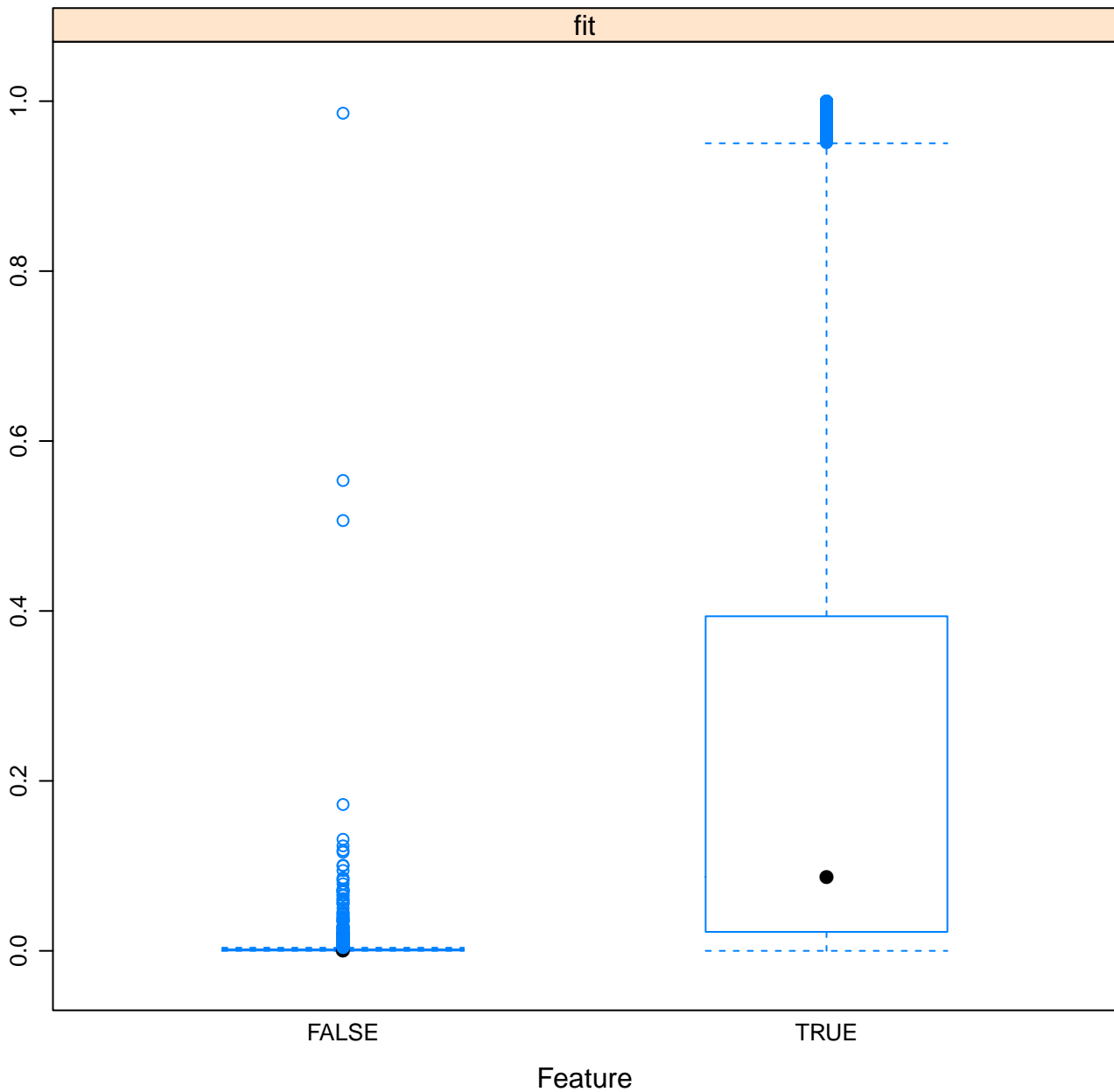
0.0

0.5

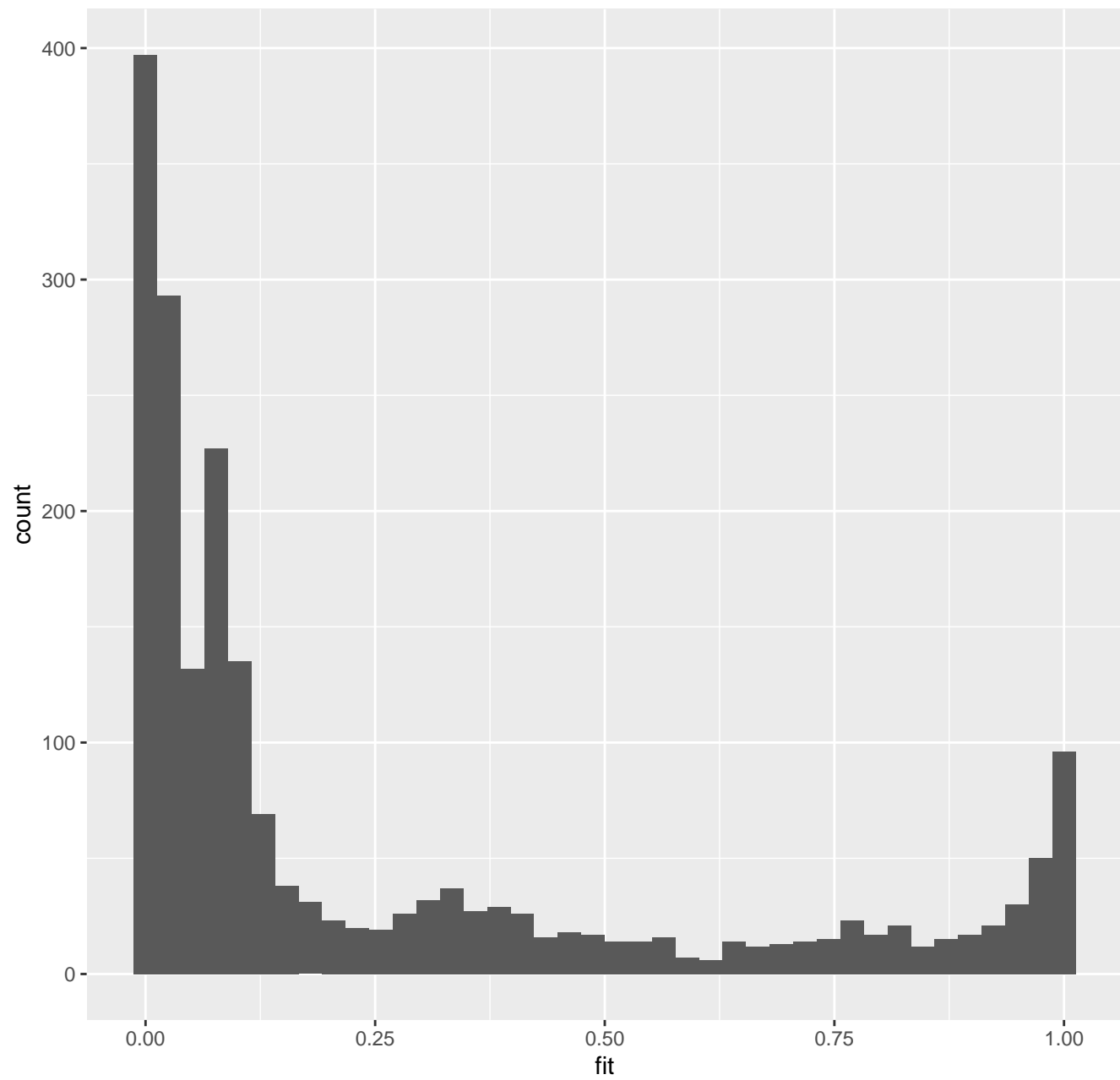
1.0

Feature

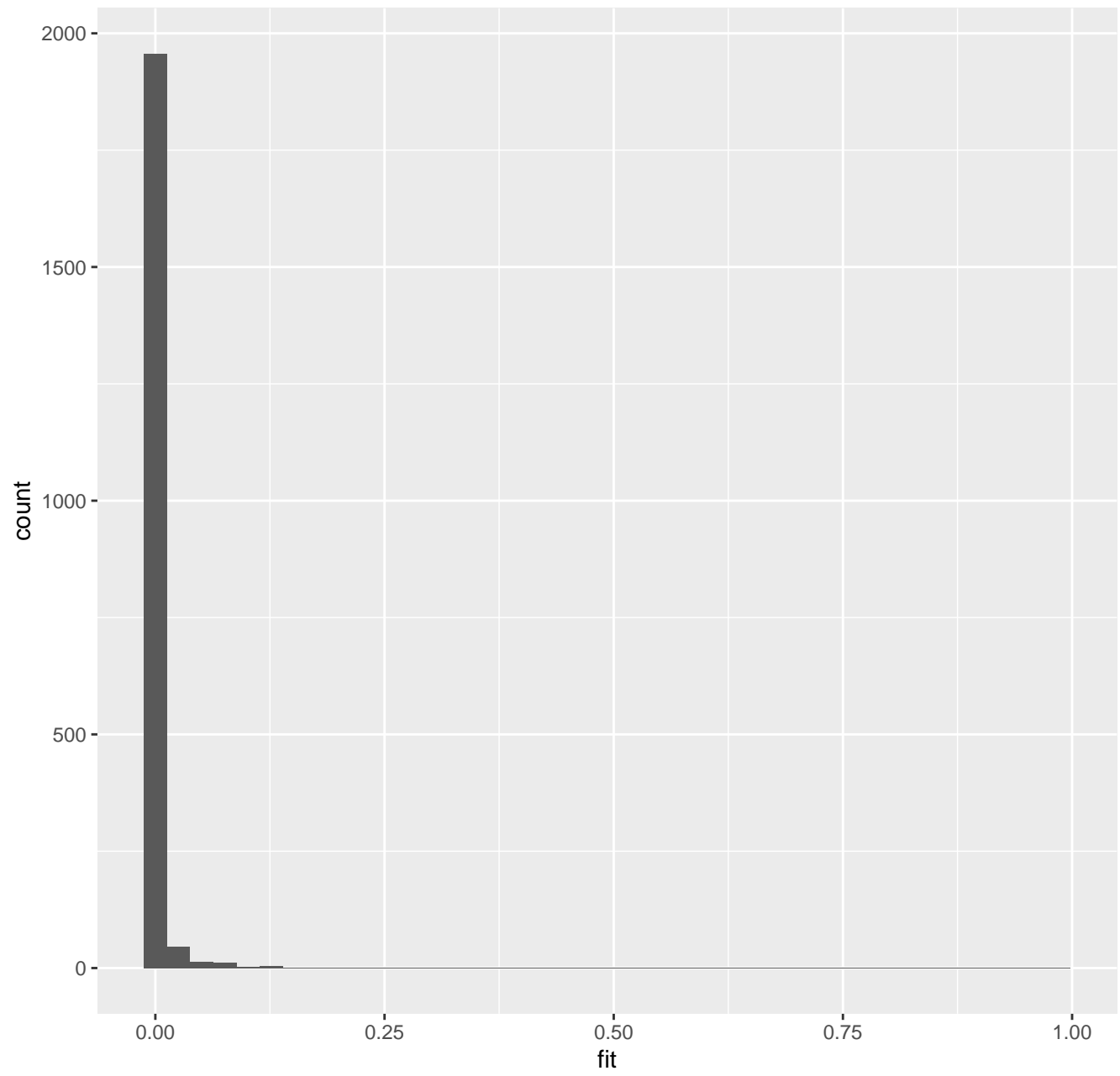




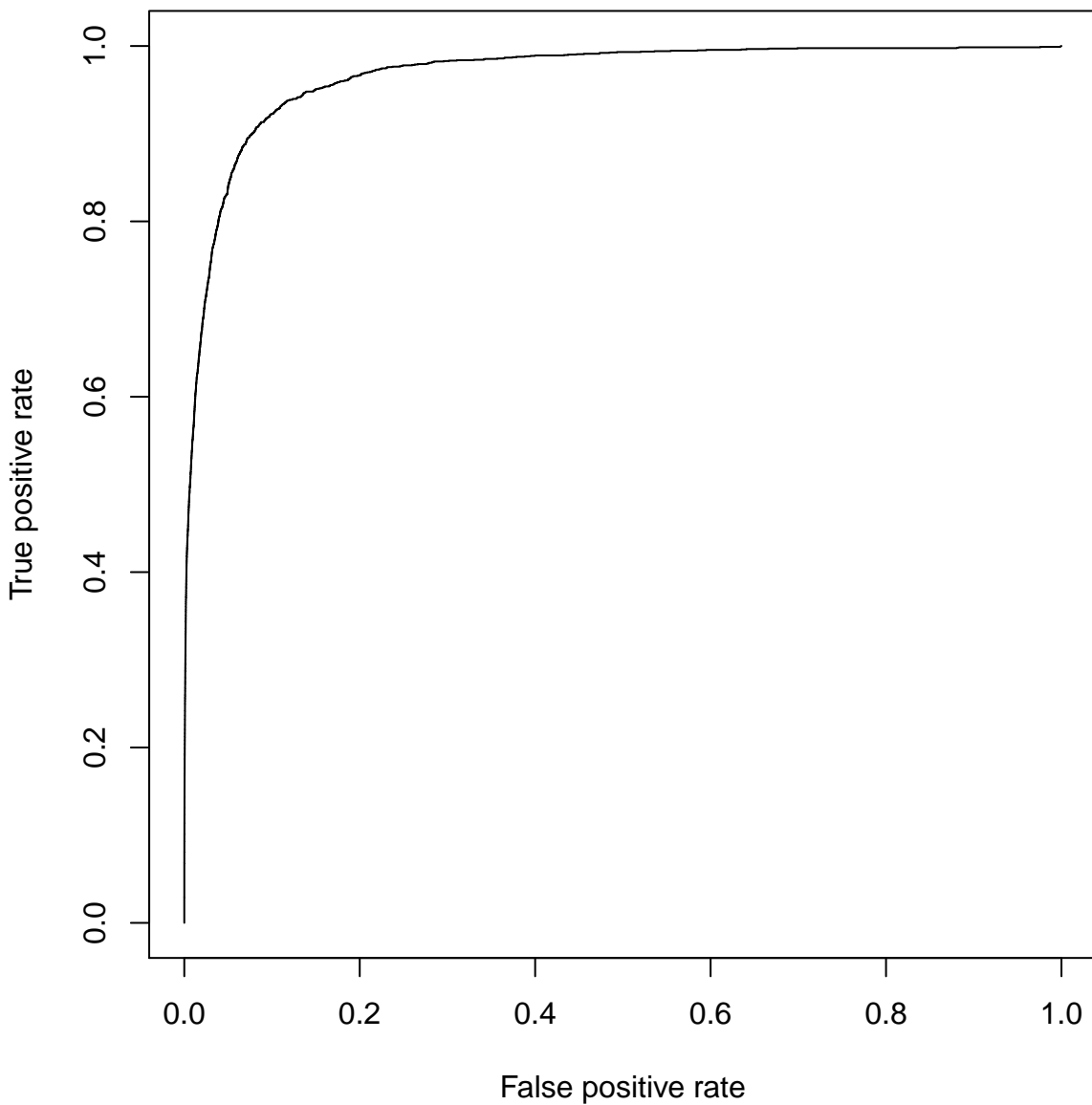
dy\_payer == TRUE



dy\_payer == TRUE



**transformed model AUC = 0.967**



```
$note  
[1] "features transformed and scaled"
```

```
$rcd_optimal_cut_off  
[1] 0.1596385
```

```
$rcd_optimal_confusion_matrix
```

	FALSE	TRUE
FALSE	396779	1179
TRUE	1182	860

```
$sensitivity  
[1] 0.4217754
```

```
$precision  
[1] 0.4211557
```

```
$relative_count_difference  
[1] 1.001471
```



FALSE



TRUE



fit

1000

800

600

400

200

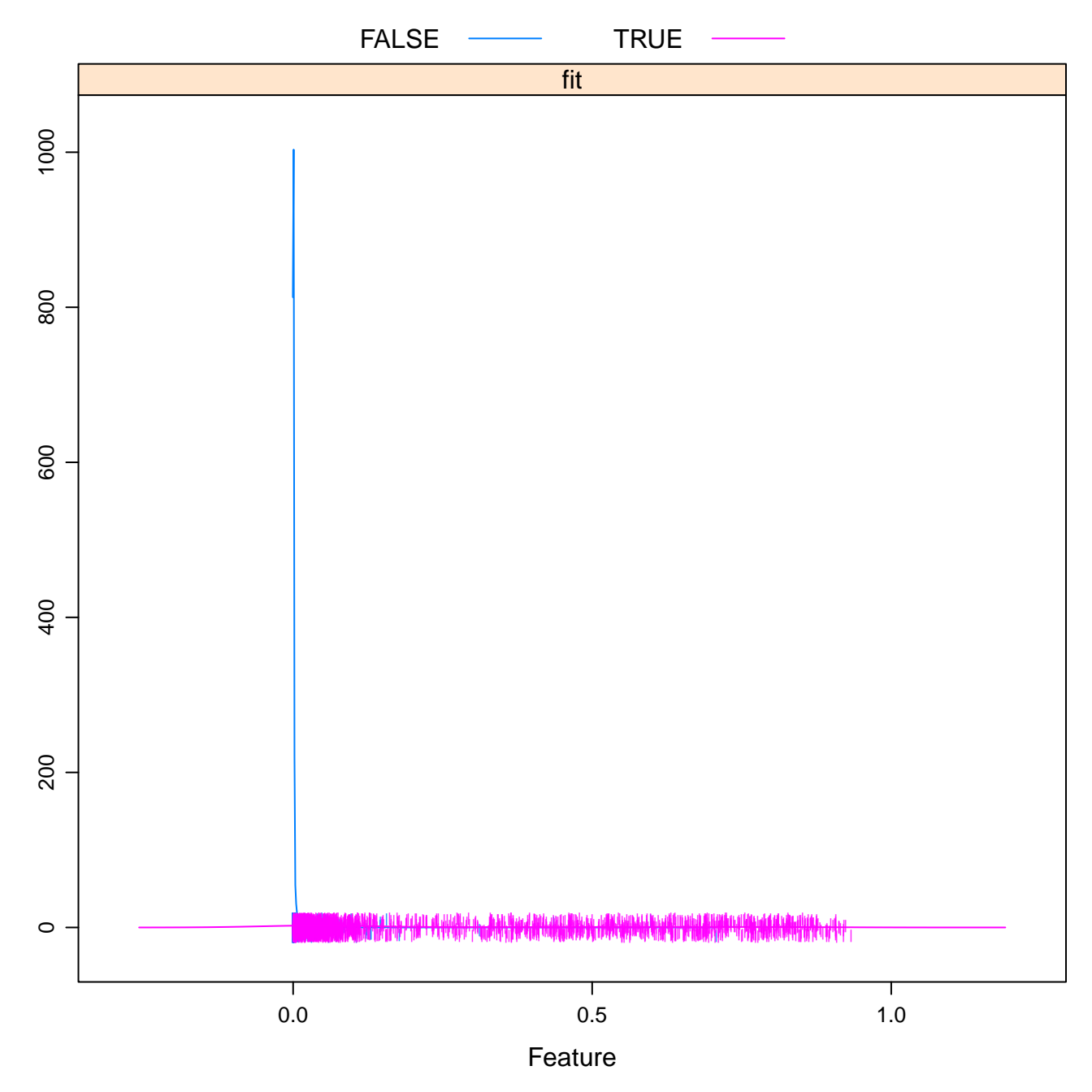
0

0.0

0.5

1.0

Feature



fit

0.8

0.6

0.4

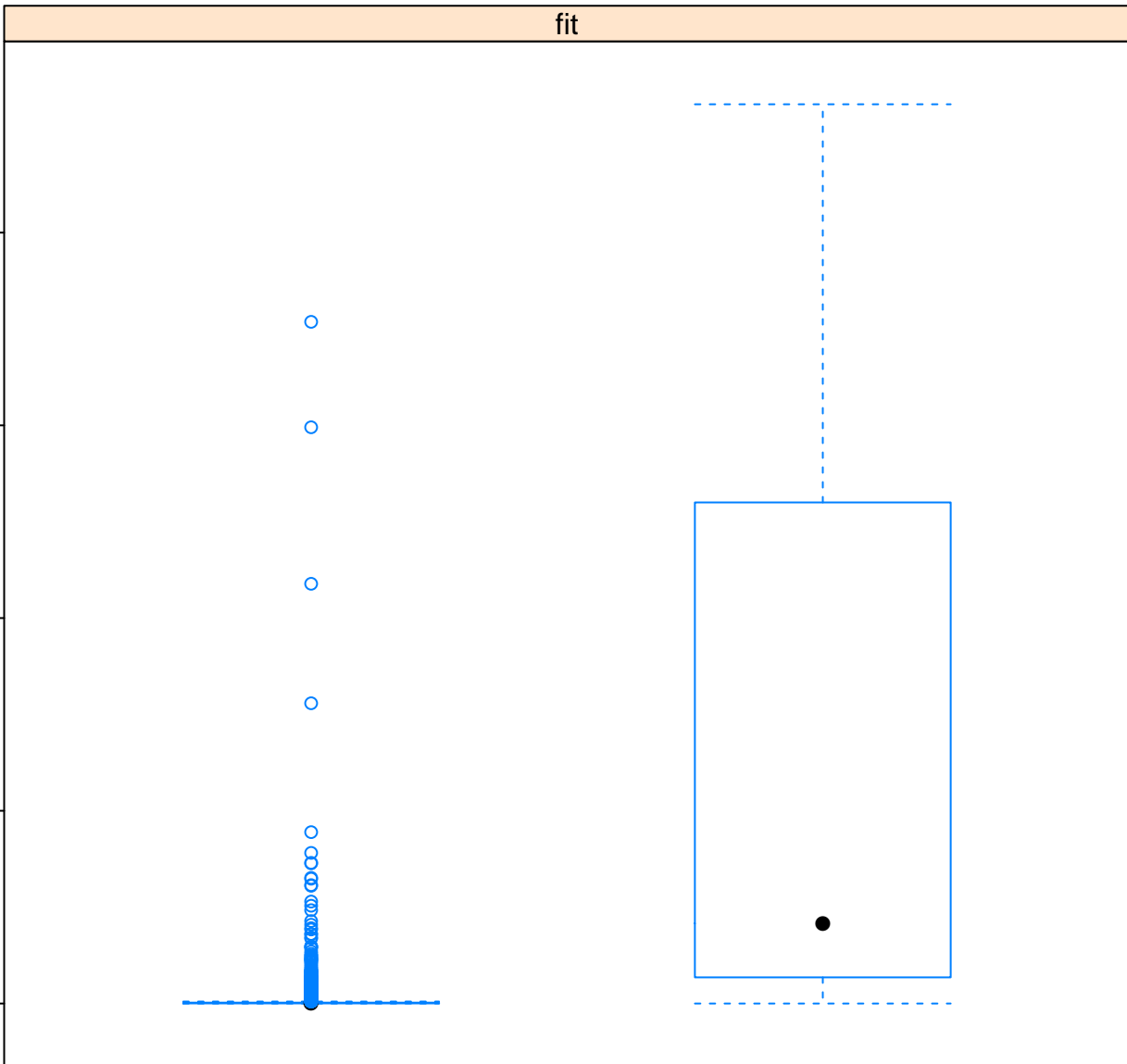
0.2

0.0

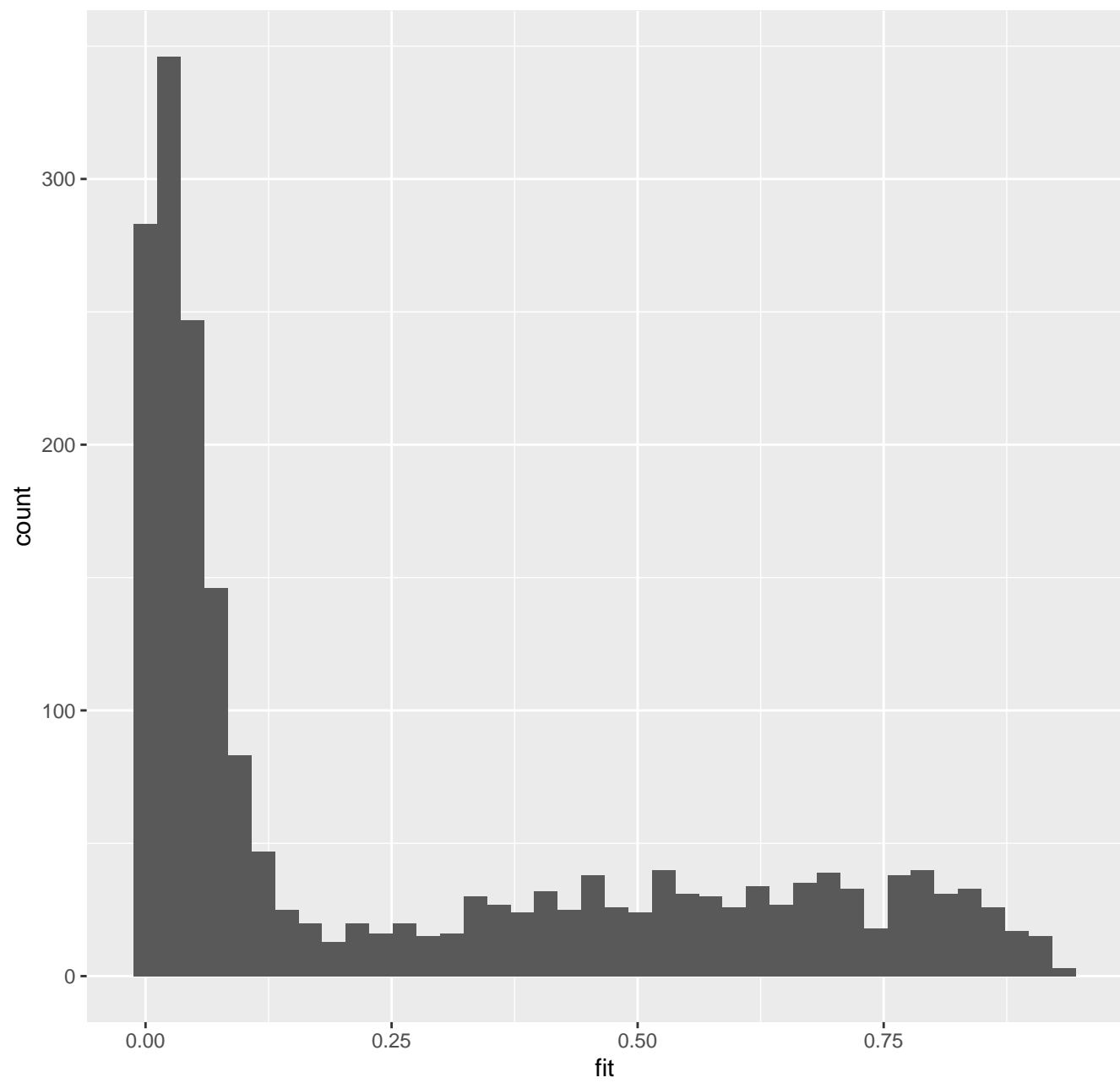
FALSE

TRUE

Feature



dy\_payer == TRUE



dy\_payer == TRUE

