

# a Bioconductor package to identify outliers in rare diseases DNA methylation data

## Supplementary material

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## Introduction

Rare diseases are pathologies with a low prevalence ( $< 1$  per 2,000 people). Most of these pathologies have an onset during childhood and a strong genetic etiology. Due to their low prevalence, there is a lack of knowledge which causes a delay in diagnosis and a lack of effective treatment. Thus, affecting the life expectancy and quality of the patient. Current methodologies for identifying mutations related to rare diseases have relied on Whole Genomic Sequencing. Despite exhaustive assessments, in a large proportion of case subjects, the underlying genetic etiology is not identified or the clinical assessment does not indicate a diagnosis. In order to overcome this difficulty, genome-wide DNA methylation analysis has been proposed to facilitate the molecular diagnosis of unresolved clinical cases and be consider for routine clinical assessment. We developed **epimutations**, a method that detects outliers in DNA methylation patterns associated with the diseases as proposed by (Aref-Eshghi et al. 2019). We validated our method by performing simulations based on the data and results obtained in the survey (Garg et al. 2020).

## Data collection

The data were obtained for the studies previously described (Garg et al. 2020). The datasets were downloaded from Gene Expression Omnibus (GEO). We accessed DNA methylation data from a total 1, 417 individuals from GSE51032 and GSE111629 cohorts. The DNA methylation profiles were generated using the Illumina 450k Human Methylation BeadChip.

The GSE51032 study analysed primary cancers samples: 424 cancer free, 235 primary breast cancer, 166 primary colorectal cancer and 20 other primary cancers. The GSE111629 cohort 335 Parkinson's disease and 237 control samples.

## Validation

We evaluated the performance of the method using TPR (True Positive Rate), False Positive Rate (FPR) and accuracy. We use the TPR to measure the proportion of detected epivariations by the **epimutations** approach present in the validated (Garg et al. 2020). FPR to calculate the identified epimutations outside the once found in (Garg et al. 2020), whether validated or not. The accuracy measures the closeness of the detected epimutation to the validated regions.

We select samples differently depending on the study group and measure to compute. Control samples were selected randomly using different sample size: 20, 30, 40, 50, 60, 70, 80, 90 and 100. However, case samples

were selected considering validated epimutations (for TPR and accuracy) or excluding epivariations found (for FPR) (Garg et al. 2020).

The validated epimutations in table 1 were only present on 5 individuals: GSM1235784 from GSE51032 cohort and GSM3035933, GSM3035791, GSM3035807 and GSM3035685 from GSE111629. Therefore, they were established as case samples when computing TPR and accuracy. Nevertheless, we compute FPR excluding the samples containing at least one epimutation found by (Garg et al. 2020). For the remaining case samples, 4 were selected randomly in each execution.

We execute 100 times the same process for each control sample size. We define for the analysis regions of  $\approx 20$  kb containing  $\geq 3$  GpGs.

Table 1: validated epimutations (Garg et al. 2020).

Chromosome	Start	End	Width	Strand	Samples
chr17	46018653	46019185	533	*	GSM1235784/GSM3035791
chr19	11199850	11200147	298	*	GSM3035685
chr5	10249760	10251253	1494	*	GSM3035933
chr5	67583971	67584381	411	*	GSM3035791/GSM3035807

Additionally, we have plotted the methylation values of the samples in the regions where the validated epimutations were found.

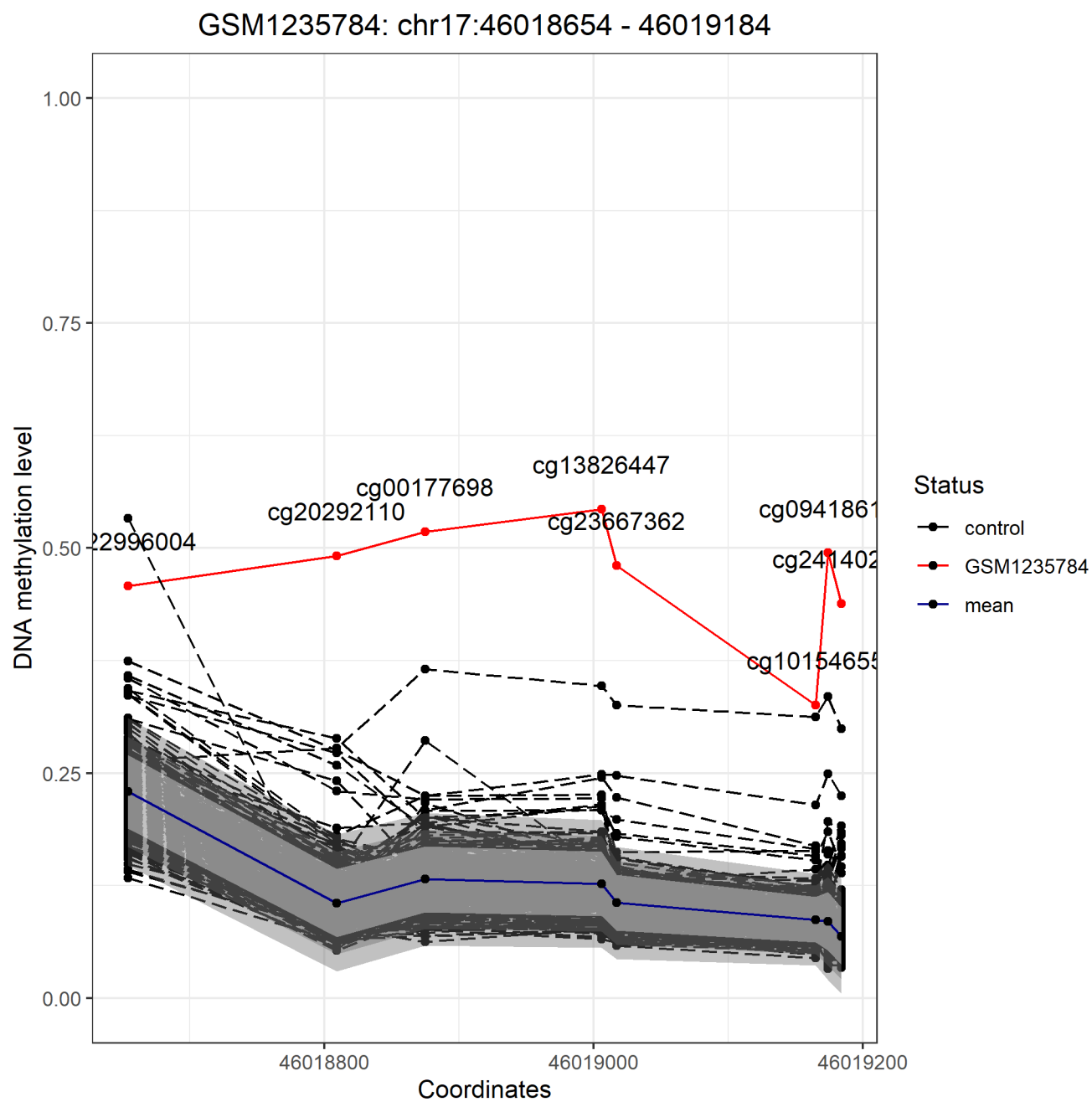


Figure 1: GSE51032 samples in the region chr17:46018654-46019184

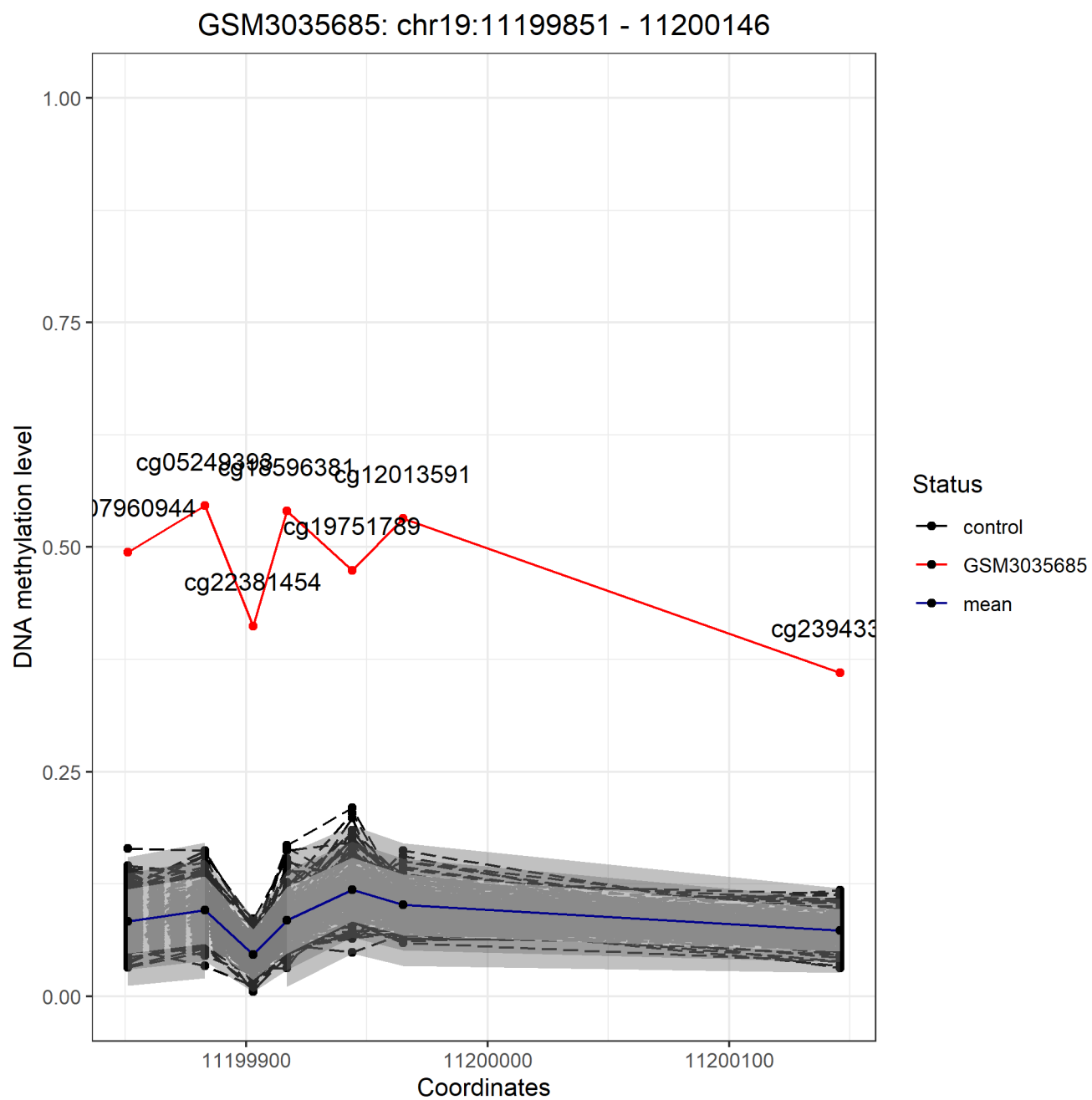


Figure 2: GSE111629 samples in the region chr19:11199851-11200146



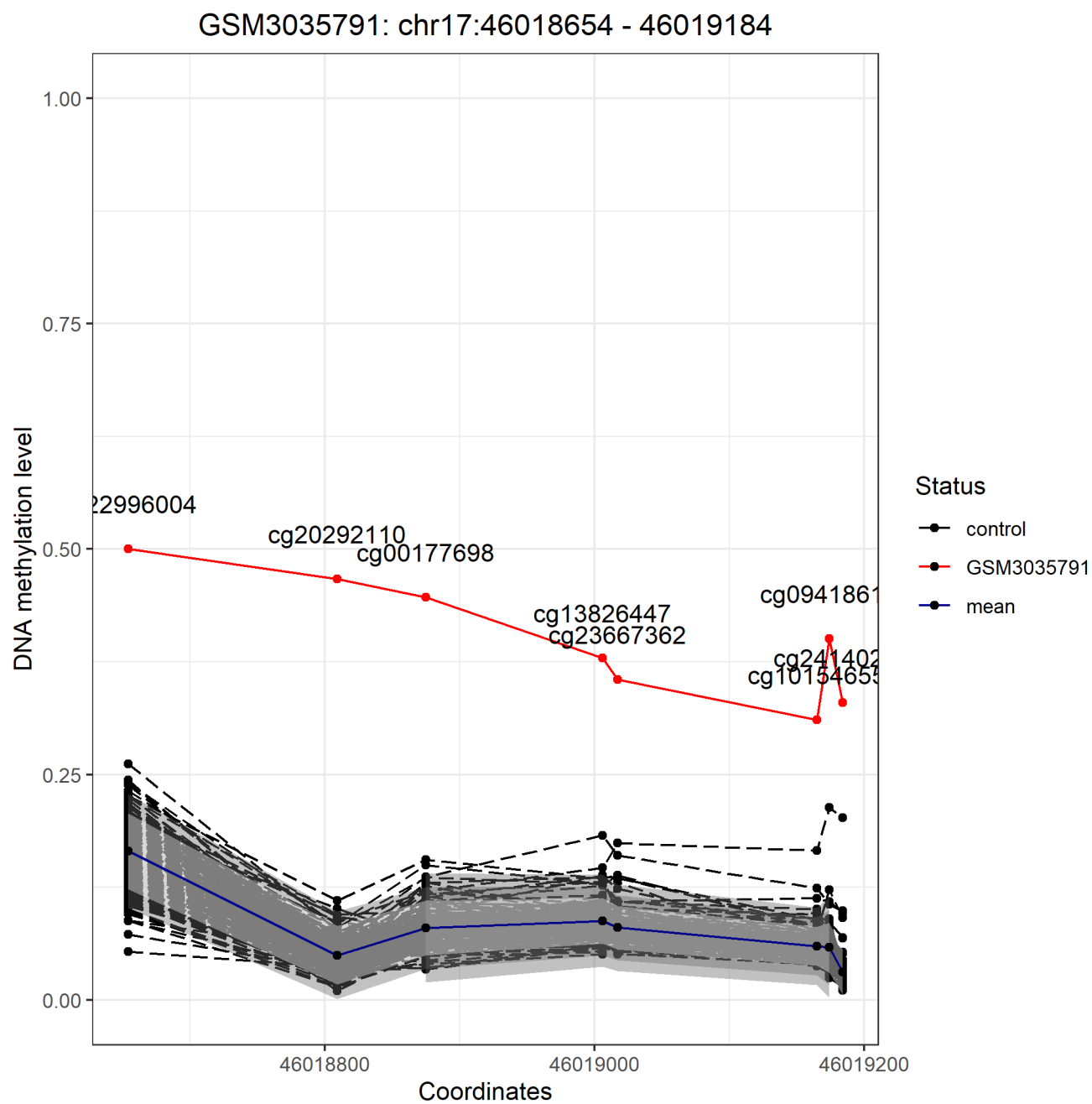


Figure 4: GSE111629 samples in the region chr17:46018654-46019184

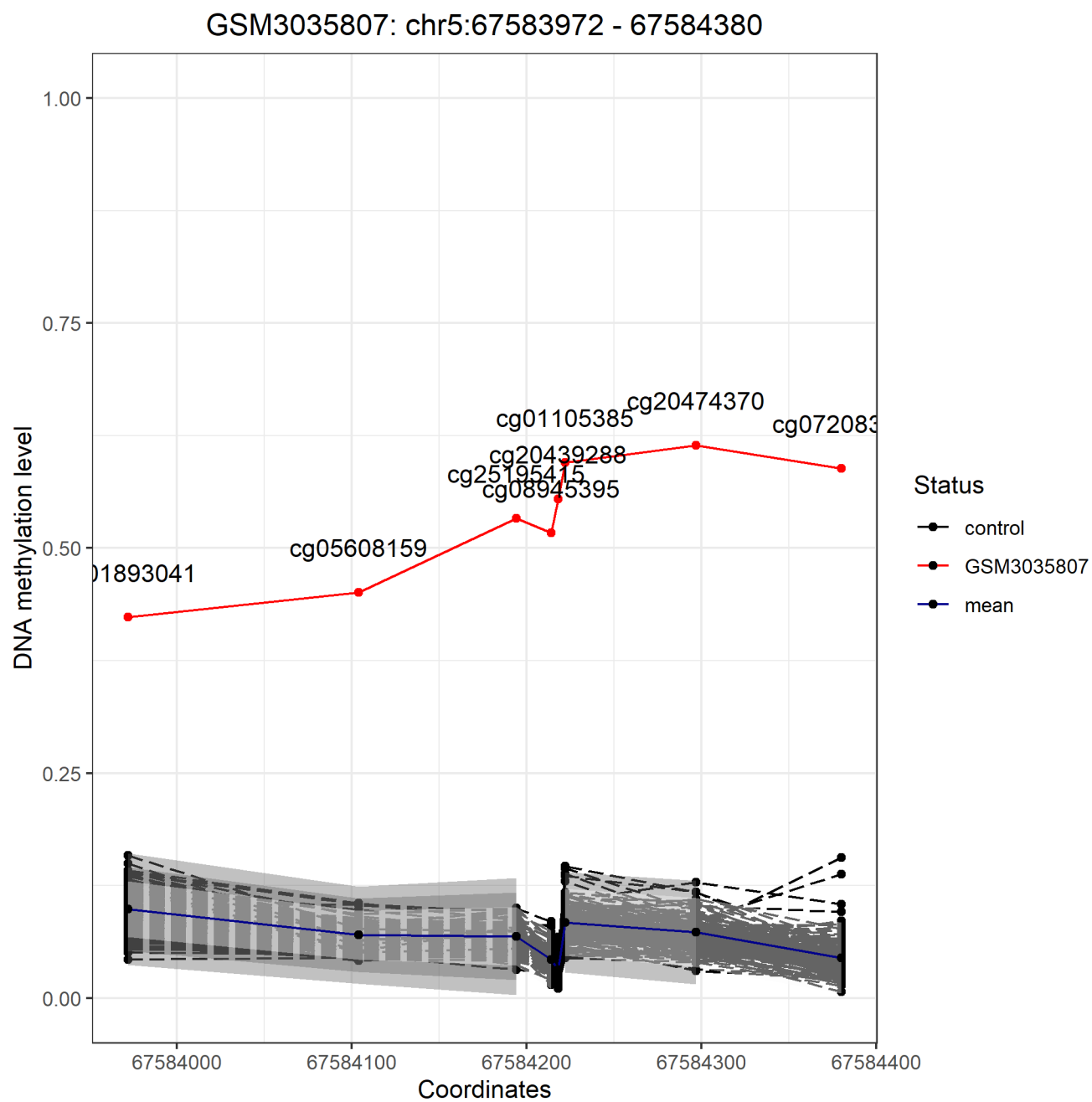


Figure 5: GSE111629 samples in the region chr5:67583972-67584380

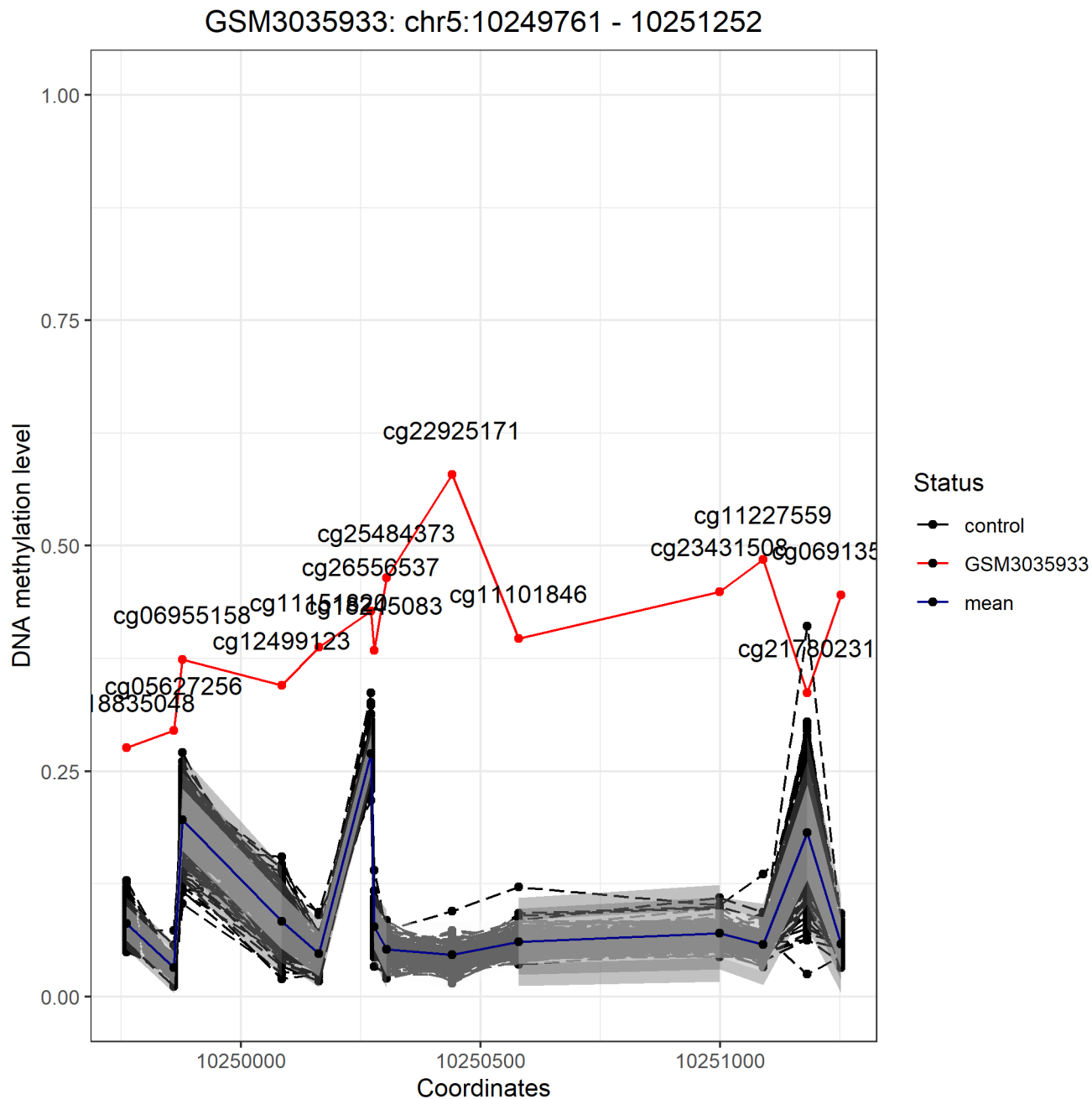


Figure 6: GSE111629 samples in the region chr5:10249761-10251252

## Results

We compare GSM1235784 case sample against randomly selected control samples from GSE51032 and GSM3035933, GSM3035791, GSM3035807 and GSM3035685 case samples against controls from GSE111629 specifying a region of 20 kb and  $\geq 3$  GpGs.

We obtained similar results in both cohorts. We observed that the methods manova, mahalanobis distance and multivariate linear models identified the validated epimutations with a TPR of  $> 99\%$  even if the control



sample is small. However, the TPR in isolation forest increases together with the number of control samples obtaining a  $TPR \geq 75$  with 50 control samples or more. The TPR in barbosa and beta approaches for GSE51032 dataset is small ( $< 50\%$ ). Nonetheless, for GSE111629 the TPR value increases considerably  $> 99\%$ . Regarding the accuracy, all the statistical approaches detect the epivariants with  $> 80\%$  of closeness to the validated epimutations.

We detected possible epivariations outside the epimutations found by (Garg et al. 2020) selecting control and case samples randomly. For the analysis, we selected regions of 20 kb and  $\geq 3$  GpGs. We compared each case sample individually against control samples. We observed that in both cohorts and for every approach the FPR value is very small  $< 0.01\%$ .

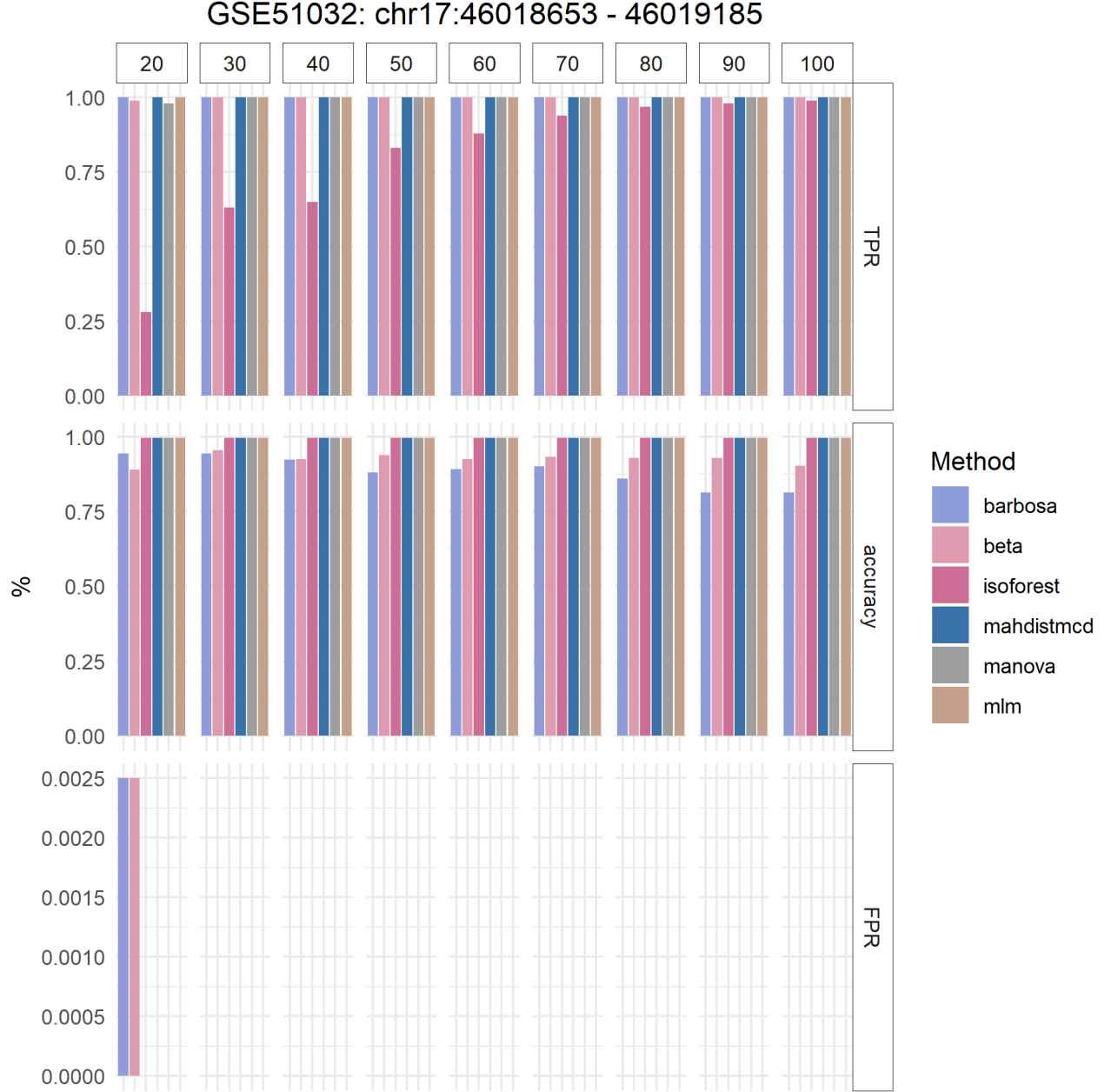


Figure 7: epimutations performance for GSE51032 cohort detecting the epivariation located in chr5:10249760-10251253

method	TPR	accuracy	FPR
<b>n20</b>			
manova	98	99.6	0.00
mlm	100	99.6	0.00
mahdistmcd	100	99.6	0.00
isoforest	28	99.6	0.00
barbosa	100	94.4	0.25
beta	99	89.0	0.25
<b>n30</b>			
manova	100	99.6	0.00
mlm	100	99.6	0.00
mahdistmcd	100	99.6	0.00
isoforest	63	99.6	0.00
barbosa	100	94.4	0.00
beta	100	95.4	0.00
<b>n40</b>			
manova	100	99.6	0.00
mlm	100	99.6	0.00
mahdistmcd	100	99.6	0.00
isoforest	65	99.6	0.00
barbosa	100	92.4	0.00
beta	100	92.6	0.00
<b>n50</b>			
manova	100	99.6	0.00
mlm	100	99.6	0.00
mahdistmcd	100	99.6	0.00
isoforest	83	99.6	0.00
barbosa	100	88.0	0.00
beta	100	93.8	0.00
<b>n60</b>			
manova	100	99.6	0.00
mlm	100	99.6	0.00
mahdistmcd	100	99.6	0.00
isoforest	88	99.6	0.00
barbosa	100	89.2	0.00
beta	100	92.6	0.00
<b>n70</b>			
manova	100	99.6	0.00
mlm	100	99.6	0.00
mahdistmcd	100	99.6	0.00
isoforest	94	99.6	0.00
barbosa	100	90.0	0.00
beta	100	93.2	0.00
<b>n80</b>			
manova	100	99.6	0.00
mlm	100	99.6	0.00
mahdistmcd	100	99.6	0.00
isoforest	97	99.6	0.00
barbosa	100	86.0	0.00
beta	100	92.9	0.00
<b>n90</b>			
manova	100	99.6	0.00
mlm	100	99.6	0.00
mahdistmcd	100	99.6	0.00
isoforest	98	99.6	0.00
barbosa	100	81.3	0.00
beta	100	92.9	0.00
<b>n100</b>			
manova	100	99.6	0.00

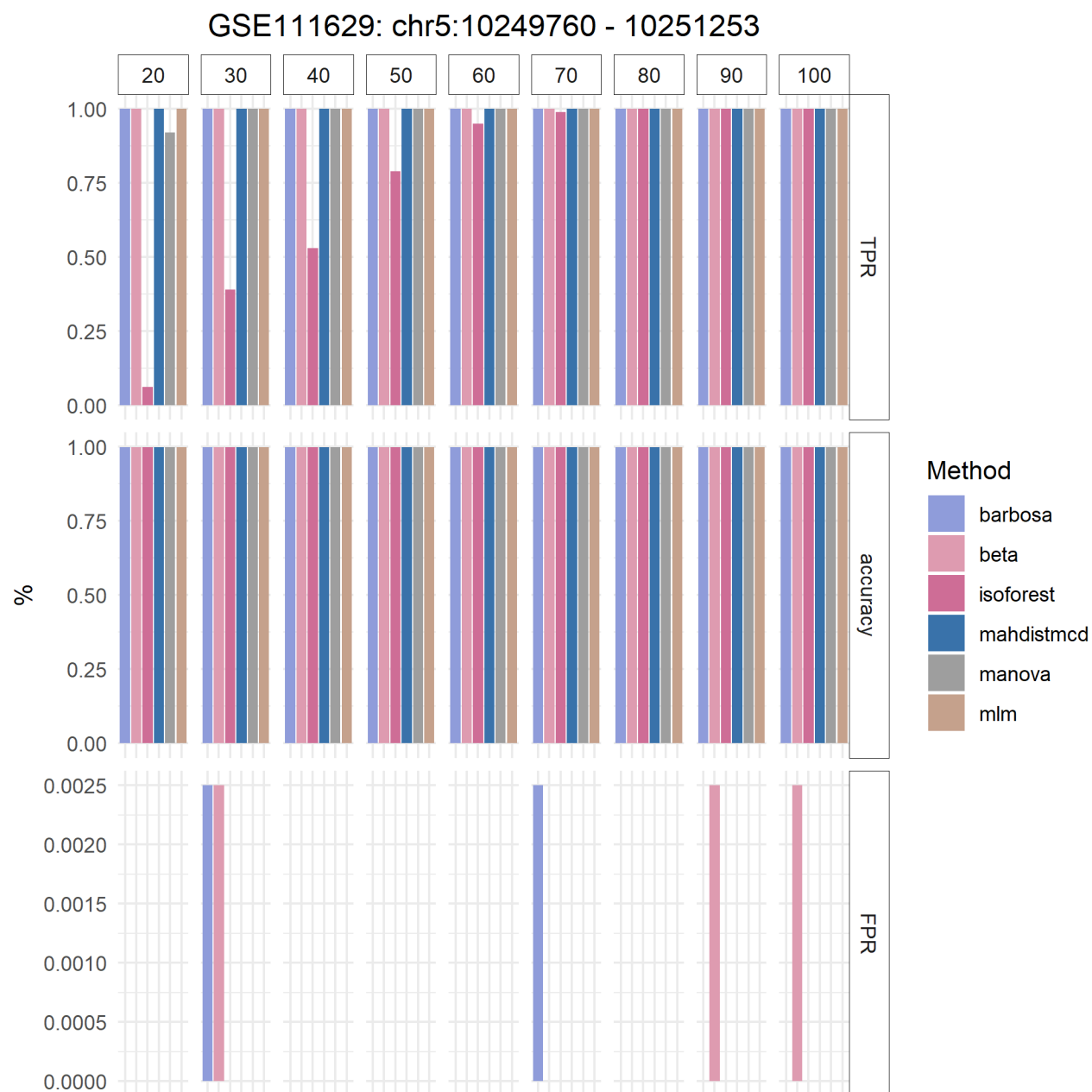


Figure 8: epimutations performance using GSE111629 cohort to detect the epivariation located in chr5:10249760-10251253

method	n	TPR	accuracy	FPR
<b>n20</b>				
barbosa	100	100.000	92.825	0.00
beta	100	100.000	92.825	0.25
isoforest	100	98.625	93.000	0.00
mahdistmcd	100	100.000	92.825	0.00
manova	100	100.000	92.825	0.00
mlm	100	100.000	92.825	0.00
<b>n30</b>				
barbosa	20	100.000	92.825	0.00
beta	20	87.500	92.800	0.00
isoforest	20	11.500	86.800	0.00
mahdistmcd	20	100.000	92.825	0.00
manova	20	98.000	92.825	0.00
mlm	20	100.000	92.850	0.00
<b>n40</b>				
barbosa	30	100.000	92.825	0.25
beta	30	87.500	92.775	0.25
isoforest	30	28.000	93.150	0.00
mahdistmcd	30	100.000	92.825	0.00
manova	30	100.000	92.825	0.00
mlm	30	100.000	92.825	0.00
<b>n50</b>				
barbosa	40	100.000	92.825	0.00
beta	40	87.500	92.800	0.00
isoforest	40	46.375	93.950	0.00
mahdistmcd	40	100.000	92.825	0.00
manova	40	100.000	92.825	0.00
mlm	40	100.000	92.825	0.00
<b>n60</b>				
barbosa	50	100.000	92.825	0.00
beta	50	87.500	92.825	0.00
isoforest	50	70.125	93.500	0.00
mahdistmcd	50	100.000	92.825	0.00
manova	50	100.000	92.825	0.00
mlm	50	100.000	92.825	0.00
<b>n70</b>				
barbosa	60	100.000	92.825	0.00
beta	60	100.000	92.825	0.00
isoforest	60	78.750	93.500	0.00
mahdistmcd	60	100.000	92.825	0.00
manova	60	100.000	92.825	0.00
mlm	60	100.000	92.825	0.00
<b>n80</b>				
barbosa	70	100.000	92.825	0.25
beta	70	100.000	92.825	0.00
isoforest	70	90.125	93.575	0.00
mahdistmcd	70	100.000	92.825	0.00
manova	70	100.000	92.825	0.00
mlm	70	100.000	92.825	0.00
<b>n90</b>				
barbosa	80	100.000	92.825	0.00
beta	80	100.000	92.825	0.00
isoforest	80	96.375	93.075	0.00
mahdistmcd	80	100.000	92.825	0.00
manova	80	100.000	92.825	0.00
mlm	80	100.000	92.825	0.00
<b>n100</b>				
barbosa	90	100.000	92.825	0.00

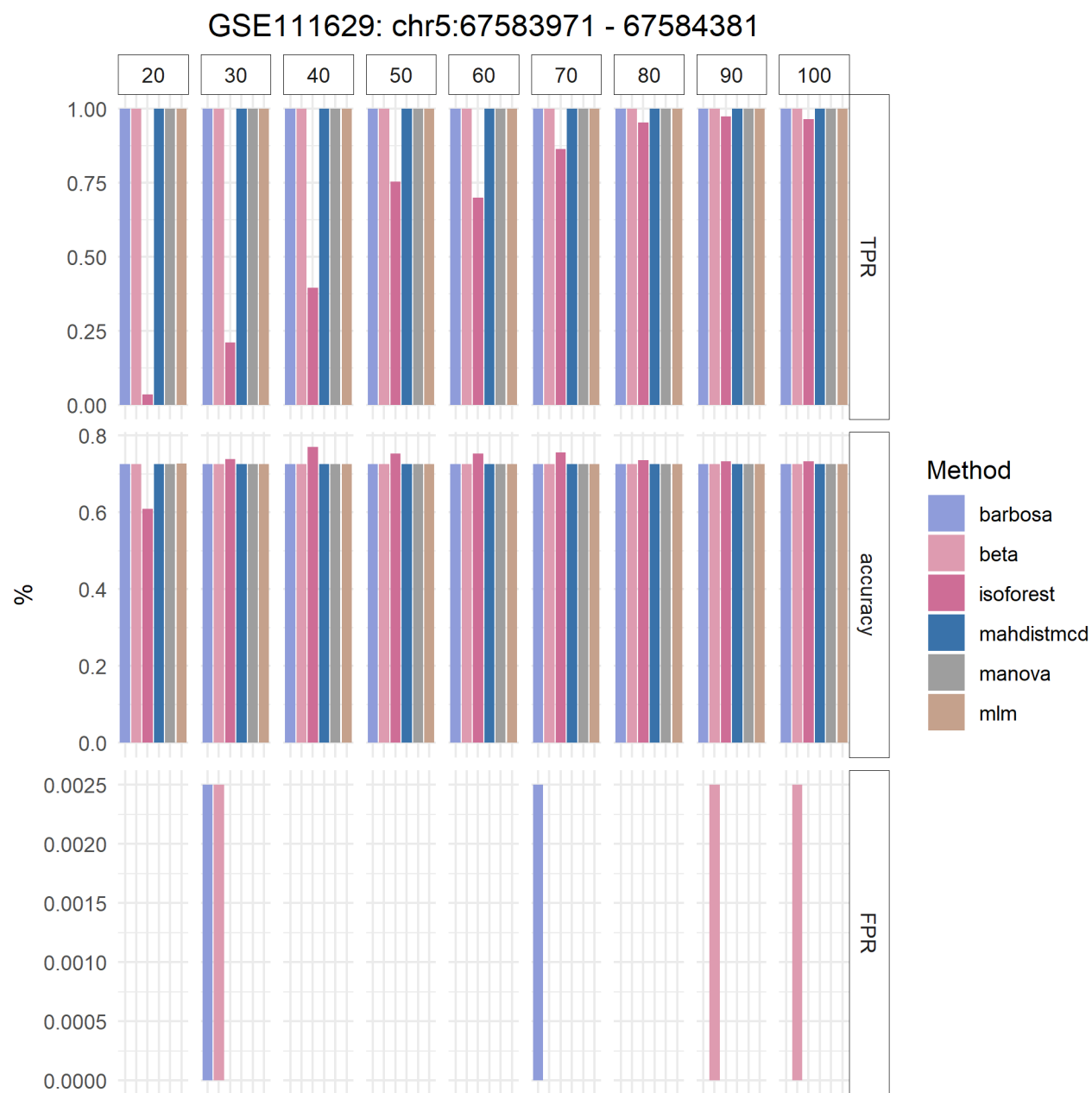


Figure 9: epimutations performance using GSE111629 cohort to detect the epivariation located in chr5:67583971-67584381

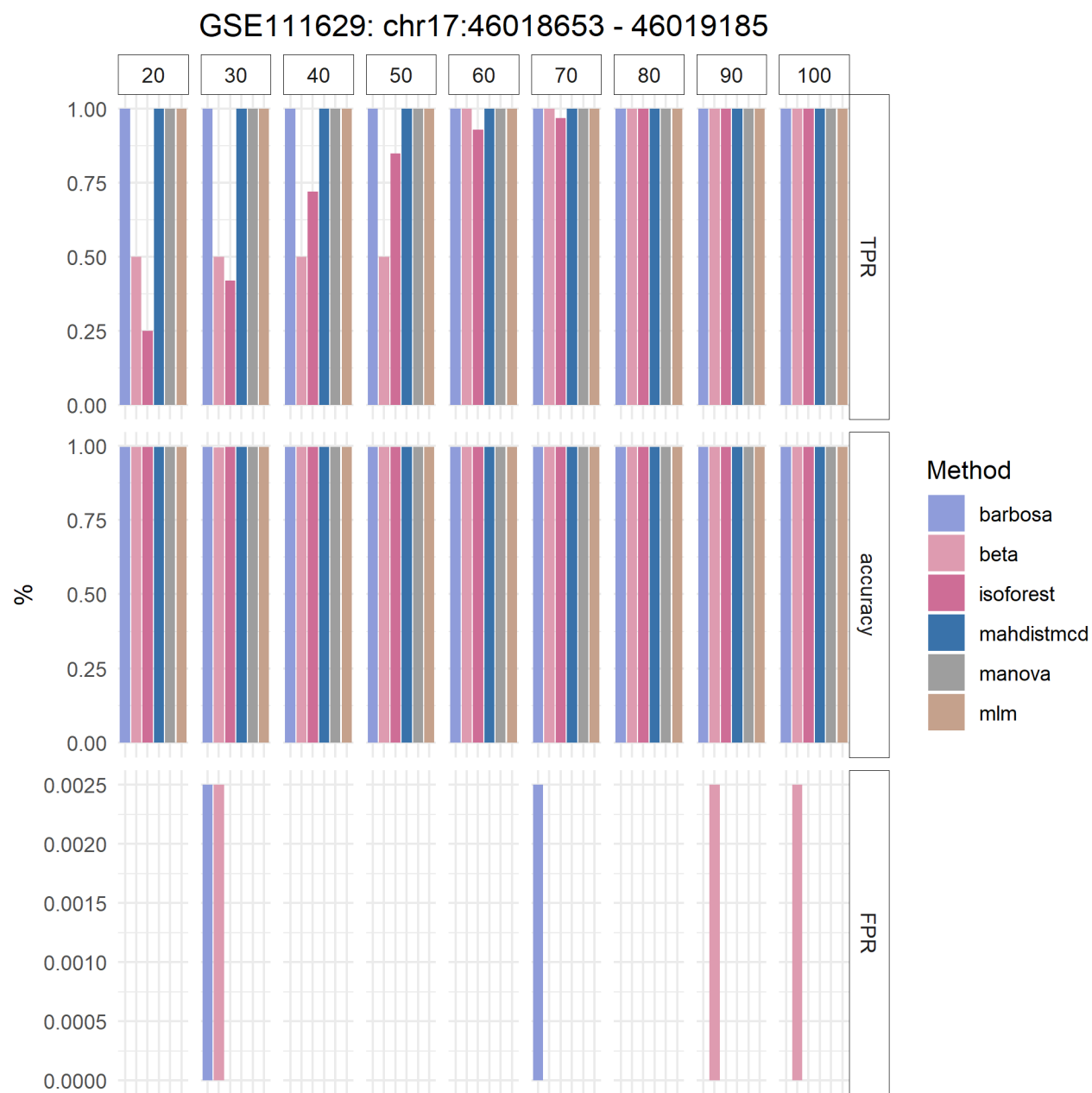


Figure 10: epimutations performance using GSE111629 cohort to detect the epivariation located in chr17:46018653-46019185

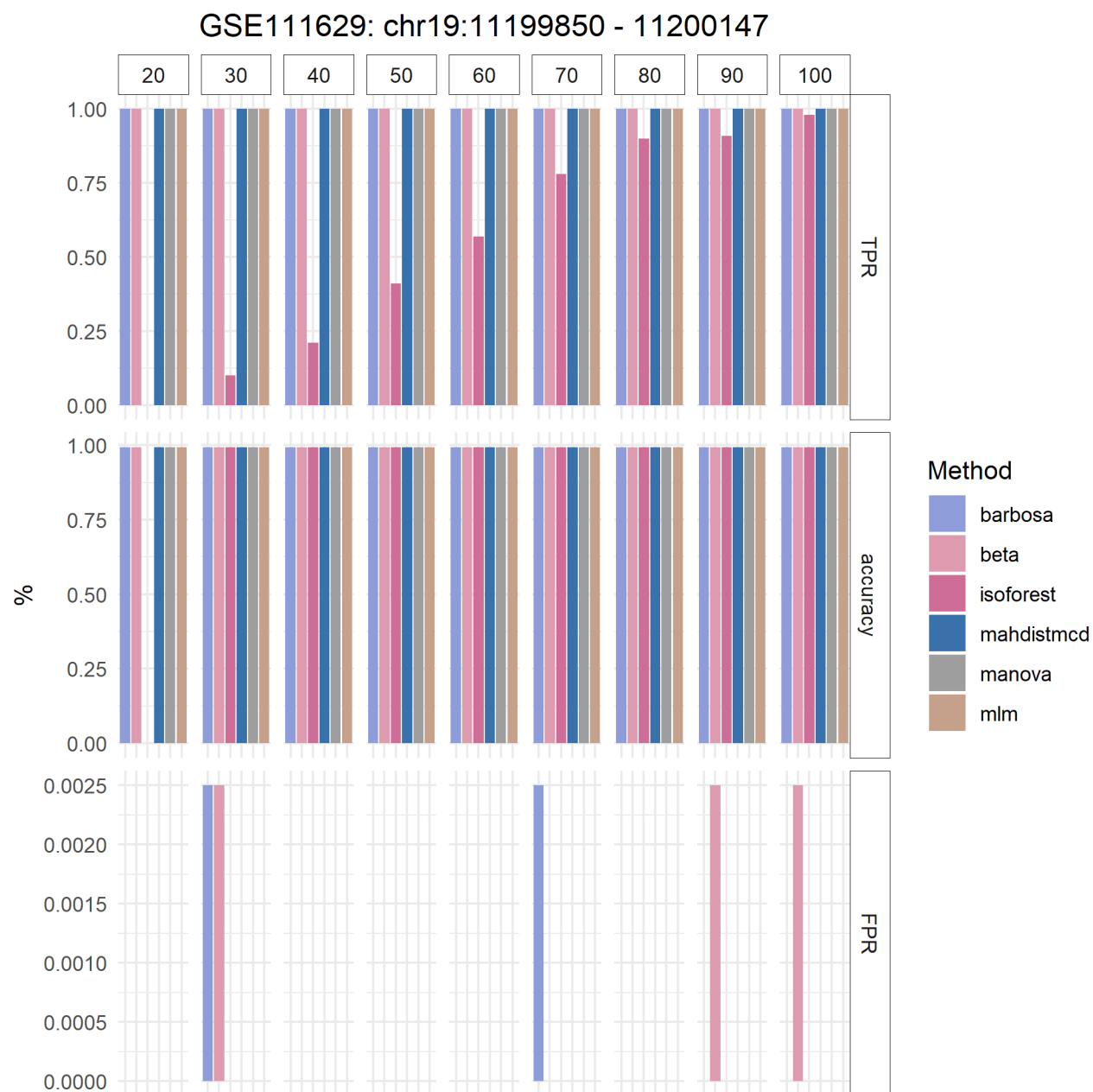


Figure 11: epimutations performance using GSE111629 cohort to detect the epivariation located in chr5:11199850-11200147

## References

Aref-Eshghi, Erfan, Eric G. Bend, Samantha Colaiacovo, Michelle Caudle, Rana Chakrabarti, Melanie Napier, Lauren Brick, et al. 2019. “Diagnostic Utility of Genome-Wide Dna Methylation Testing in Genetically Unsolved Individuals with Suspected Hereditary Conditions.” *The American Journal of Human Genetics*. <https://doi.org/https://doi.org/10.1016/j.ajhg.2019.03.008>.

Garg, Paras, Bharati Jadhav, Oscar L Rodriguez, Nihir Patel, Alejandro Martin-Trujillo, Miten Jain, Sofie Metsu, et al. 2020. “A Survey of Rare Epigenetic Variation in 23,116 Human Genomes Identifies Disease-Relevant Epivariations and Cgg Expansions.” *The American Journal of Human Genetics* 107 (4): 654–69.