

Algorithms for Automatized Detection of Hook Effect-bearing Amplification Curves

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1 Abstract

This is a supplemental document for the study *Algorithms for Automatized Detection of Hook Effect-bearing Amplification Curves*. Quantitative real-time PCR (qPCR) is a widely used method for gene expression analysis, forensics and medical diagnostics (Dvinge and Bertone 2009; Martins et al. 2015; Sauer, Reinke, and Courts 2016).

Numerous algorithms have been developed to extract features from amplification curves such as the cycle of quantification and the amplification efficiency (Ruijter et al. 2013). There is an agreement, that these algorithms need to be evaluated and benchmarked for their performance (Kemperman and McCall 2017). But at an earlier level it is important to have a solid foundation for the data preprocessing (Spiess et al. 2015, 2016; Ronde et al. 2017). Digitalization of processes holds the promise that potential human mistakes can be spotted and that diagnostic processes can be automatized.

The aim of the study is to provide software tools and algorithms, which assists qPCR users during the analysis and quality management of their data. In particular, this study shows how it is possible to automatically detect hook effects (see Barratt and Mackay (2002)) or hook effect-like curvatures.

2 Introduction

The functions and data presented in the paper are available from <https://github.com/devSJR/PCRedux>. The data, including the RDML file, are part of the **PCRedux** package and are made available in the CSV or RDML format (Rödiger et al. 2017) for vendors independent analysis.

All analyses were implemented and conducted with the **R** statistical computing language (R Core Team 2017; Rödiger et al. 2015) and dedicated integrated development environments such as **RKward** (Rödiger et al. 2012). Further documentation can be found in the help files of the **R** packages.

3 Installation

The **hookreg()** and **hookregNL()** functions are part of the **PCRedux** package for the **R** statistical computing language. Download from CRAN <http://cran.r-project.org/> the **R** version for the required operating system and install **R**. Then start **R** and type in the prompt:

```
# Select your local mirror
install.packages("PCRedux")
```

The **PCRedux** package should just install. If this fails make sure you have write access to the destination directory and follow the instructions of the **R** documentation:

```
# The following command points to the help for download and install of packages
# from CRAN-like repositories or from local files.
?install.packages()
```

The package can be installed as the latest development version using the devtools **R** package.

```
# Install devtools, if you haven't already.
install.packages("devtools")

library(devtools)
install_github("devSJR/PCRedux")
```

It is recommended to use software with an integrated development environment such as **RKward** (Rödiger et al. 2012). To work with RDML data it is recommend to use the **RDML** package (\geq v.0.9-9) by invoking the **rdmlEdit()** function (for details see Rödiger et al. (2017)) or the **rdmlEdit** GUI web server (section ??). The RDML file **hookreg.rdml** contains the amplification curve data. However, other software package (e.g., (Lefever et al. 2009; Ruijter et al. 2015)) can also be used to work with the RDML data file format.

4 Results for the analysis of the `hookreg.rdml` data set by `human-rater()`

All calculations in the following sections were employed on the `hookreg.rdml` data **R** environment by the **RDML** package (Rödiger et al. 2017). An overview of the used samples and the qPCR detection chemistries and the classification by two humans (“Hook effect-like Rater 1”, “Hook effect-like Rater 2”) is shown in Table 1.

Loading experiment: `exp1` run: `run1`

All amplification curves were plotted according to their experiment conditions. They differed in the target molecules (e.g., *MLC-2v*, *BRCA1*) and the detection chemistries (e.g., EvaGreen, SybrGreen, hydrolysis probes). Figure 1 shows seven plots for the corresponding experiments. The amplification curves were not preprocessed to preserve the curvature. Selected amplification curves were noisy (e.g., Figure 1F), had overshoots or undershoot in the background phase (e.g., Figure 1E-G), a short hook phase (e.g., Figure 1D). Amplification curves of Figure 1A, D, F and F exhibited a clearly visible hook effect or a hook like effect.

```
par(mfrow=c(4,2))

# Plot all data of the hookreg.rdml-file according to their type.
# Synthetic template, detected with Syto-13
matplot(data[, 1], data[, 2:13], type="l", lty=1, lwd=2, ylab="RFU", xlab="Cycle")
mtext("A", cex = 1.8, side = 3, adj = 0, font = 2)

# Human MLC-2v, detected with a hydrolysis probe.
matplot(data[, 1], data[, 14:45], type="l", lty=1, lwd=2, ylab="RFU", xlab="Cycle")
mtext("B", cex = 1.8, side = 3, adj = 0, font = 2)

# S27a housekeeping gene, detected with SybrGreen I.
matplot(data[, 1], data[, 46:69], type="l", lty=1, lwd=2, ylab="RFU", xlab="Cycle")
mtext("C", cex = 1.8, side = 3, adj = 0, font = 2)

# Whole genome amplification, detected with EvaGreen.
matplot(data[, 1], data[, 70:71], type="l", lty=1, lwd=2, ylab="RFU", xlab="Cycle")
mtext("D", cex = 1.8, side = 3, adj = 0, font = 2)

# Human BRCA1 gene, detected with a hydrolysis probe.
matplot(data[, 1], data[, 72:87], type="l", lty=1, lwd=2, ylab="RFU", xlab="Cycle")
mtext("E", cex = 1.8, side = 3, adj = 0, font = 2)

# Human NRAS gene, detected with a hydrolysis probe.
matplot(data[, 1], data[, 88:95], type="l", lty=1, lwd=2, ylab="RFU", xlab="Cycle")
mtext("F", cex = 1.8, side = 3, adj = 0, font = 2)

# Water control, detected with a hydrolysis probe.
matplot(data[, 1], data[, 96:97], type="l", lty=1, lwd=2, ylab="RFU", xlab="Cycle")
mtext("G", cex = 1.8, side = 3, adj = 0, font = 2)
```

Printout of all measured samples, their rating by two humans (rater 1 and rater 2) with their dichotomous ratings (0, no hook; 1, hook) and their sources.

- The `boggy` data (`qpcR::boggy`) set was taken from the `qpcR` package (Ritz and Spiess 2008; Spiess, Feig, and Ritz 2008).
- The `C127EGHP` data (`chipPCR::C127EGHP`) set was taken from the `chipPCR` package (Rödiger, Burdukiewicz, and Schierack 2015).
- The `testdat` data (`qpcR::testdat`) set was taken from the `qpcR` package (Ritz and Spiess 2008; Spiess,

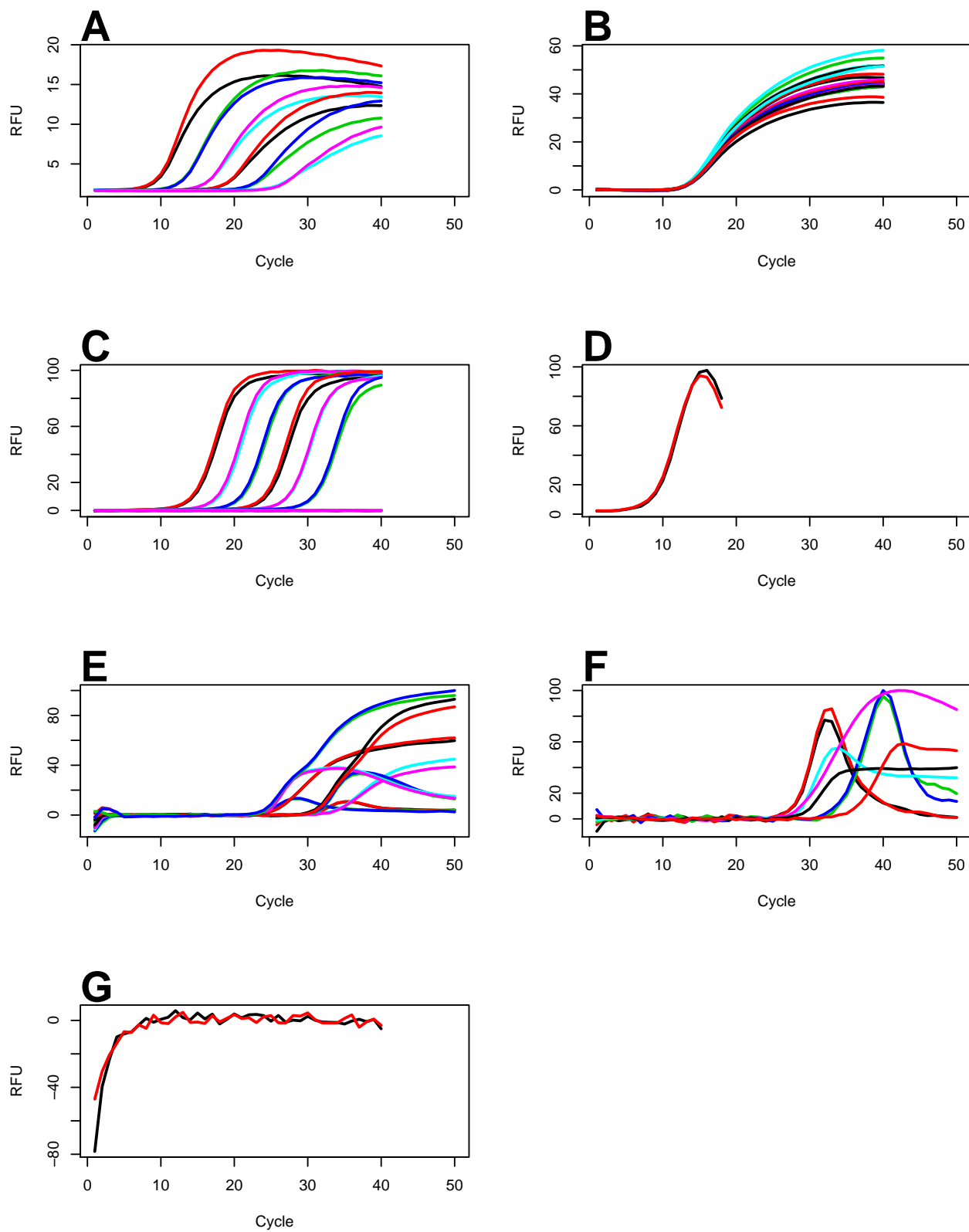


Figure 1: Amplification curves. A) Synthetic template, detected with Syto-13. B) Human *MLC-2v*, detected with a hydrolysis probe. C) *S27a* housekeeping gene, detected with SybrGreen I. D) Whole genome amplification, detected with EvaGreen. E) Human *BRCA1* gene, detected with a hydrolysis probe. F) Human *NRAS* gene, detected with a hydrolysis probe. G) Water control, detected with a hydrolysis probe. See Table 1 for details. RFU, relative fluorescence units.

Feig, and Ritz 2008).

- Other data were prepared by Evrogen laboratory experiments.

Table 1: Overview of the used amplification curve data. The samples names, data source (origin of data either from an existing data set or prepared for this study), the detection chemistries (intercalator (Syto-13, SyberGreenI, EvaGreen), hydrolysis probes (TaqMan (Cy5/BHQ2) , TaqMan (HEX/BHQ1))) and calculations by tow humans.

#	Sample	Data Source	Target	Chemistry	Hook effect-like Rater 1	Hook effect-like Rater 2	Rating Conformity
1	F1.1	qpcR-boggy	synthetic template	Syto-13	1	1	1
2	F1.2	qpcR-boggy	synthetic template	Syto-13	1	1	1
3	F2.1	qpcR-boggy	synthetic template	Syto-13	1	1	1
4	F2.2	qpcR-boggy	synthetic template	Syto-13	1	1	1
5	F3.1	qpcR-boggy	synthetic template	Syto-13	0	0	1
6	F3.2	qpcR-boggy	synthetic template	Syto-13	0	0	1
7	F4.1	qpcR-boggy	synthetic template	Syto-13	0	0	1
8	F4.2	qpcR-boggy	synthetic template	Syto-13	0	0	1
9	F5.1	qpcR-boggy	synthetic template	Syto-13	0	0	1
10	F5.2	qpcR-boggy	synthetic template	Syto-13	0	0	1
11	F6.1	qpcR-boggy	synthetic template	Syto-13	0	0	1
12	F6.2	qpcR-boggy	synthetic template	Syto-13	0	0	1
13	HP1	chipPCR-C127EGHP	MLC-2v	TaqMan (Cy5/BHQ2)	0	0	1
14	HP2	chipPCR-C127EGHP	MLC-2v	TaqMan (Cy5/BHQ2)	0	0	1
15	HP3	chipPCR-C127EGHP	MLC-2v	TaqMan (Cy5/BHQ2)	0	0	1
16	HP4	chipPCR-C127EGHP	MLC-2v	TaqMan (Cy5/BHQ2)	0	0	1
17	HP5	chipPCR-C127EGHP	MLC-2v	TaqMan (Cy5/BHQ2)	0	0	1
18	HP6	chipPCR-C127EGHP	MLC-2v	TaqMan (Cy5/BHQ2)	0	0	1
19	HP7	chipPCR-C127EGHP	MLC-2v	TaqMan (Cy5/BHQ2)	0	0	1
20	HP8	chipPCR-C127EGHP	MLC-2v	TaqMan (Cy5/BHQ2)	0	0	1
21	HP9	chipPCR-C127EGHP	MLC-2v	TaqMan (Cy5/BHQ2)	0	0	1
22	HP10	chipPCR-C127EGHP	MLC-2v	TaqMan (Cy5/BHQ2)	0	0	1
23	HP11	chipPCR-C127EGHP	MLC-2v	TaqMan (Cy5/BHQ2)	0	0	1
24	HP12	chipPCR-C127EGHP	MLC-2v	TaqMan (Cy5/BHQ2)	0	0	1
25	HP13	chipPCR-C127EGHP	MLC-2v	TaqMan (Cy5/BHQ2)	0	0	1
26	HP14	chipPCR-C127EGHP	MLC-2v	TaqMan (Cy5/BHQ2)	0	0	1
27	HP15	chipPCR-C127EGHP	MLC-2v	TaqMan (Cy5/BHQ2)	0	0	1
28	HP16	chipPCR-C127EGHP	MLC-2v	TaqMan (Cy5/BHQ2)	0	0	1
29	HP17	chipPCR-C127EGHP	MLC-2v	TaqMan (Cy5/BHQ2)	0	0	1
30	HP18	chipPCR-C127EGHP	MLC-2v	TaqMan (Cy5/BHQ2)	0	0	1
31	HP19	chipPCR-C127EGHP	MLC-2v	TaqMan (Cy5/BHQ2)	0	0	1
32	HP20	chipPCR-C127EGHP	MLC-2v	TaqMan (Cy5/BHQ2)	0	0	1
33	HP21	chipPCR-C127EGHP	MLC-2v	TaqMan (Cy5/BHQ2)	0	0	1
34	HP22	chipPCR-C127EGHP	MLC-2v	TaqMan (Cy5/BHQ2)	0	0	1
35	HP23	chipPCR-C127EGHP	MLC-2v	TaqMan (Cy5/BHQ2)	0	0	1
36	HP24	chipPCR-C127EGHP	MLC-2v	TaqMan (Cy5/BHQ2)	0	0	1
37	HP25	chipPCR-C127EGHP	MLC-2v	TaqMan (Cy5/BHQ2)	0	0	1
38	HP26	chipPCR-C127EGHP	MLC-2v	TaqMan (Cy5/BHQ2)	0	0	1
39	HP27	chipPCR-C127EGHP	MLC-2v	TaqMan (Cy5/BHQ2)	0	0	1
40	HP28	chipPCR-C127EGHP	MLC-2v	TaqMan (Cy5/BHQ2)	0	0	1
41	HP29	chipPCR-C127EGHP	MLC-2v	TaqMan (Cy5/BHQ2)	0	0	1
42	HP30	chipPCR-C127EGHP	MLC-2v	TaqMan (Cy5/BHQ2)	0	0	1
43	HP31	chipPCR-C127EGHP	MLC-2v	TaqMan (Cy5/BHQ2)	0	0	1
44	HP32	chipPCR-C127EGHP	MLC-2v	TaqMan (Cy5/BHQ2)	0	0	1
45	F1.1_td	qpcR:testedat	S27a housekeeping gene	SybrGreen I	0	0	1
46	F1.2_td	qpcR:testedat	S27a housekeeping gene	SybrGreen I	0	0	1
47	F1.3_td	qpcR:testedat	S27a housekeeping gene	SybrGreen I	0	0	1
48	F1.4_td	qpcR:testedat	S27a housekeeping gene	SybrGreen I	0	0	1
49	F2.1_td	qpcR:testedat	S27a housekeeping gene	SybrGreen I	0	0	1
50	F2.2_td	qpcR:testedat	S27a housekeeping gene	SybrGreen I	0	0	1
51	F2.3_td	qpcR:testedat	S27a housekeeping gene	SybrGreen I	0	0	1
52	F2.4_td	qpcR:testedat	S27a housekeeping gene	SybrGreen I	0	0	1
53	F3.1_td	qpcR:testedat	S27a housekeeping gene	SybrGreen I	0	0	1
54	F3.2_td	qpcR:testedat	S27a housekeeping gene	SybrGreen I	0	0	1
55	F3.3_td	qpcR:testedat	S27a housekeeping gene	SybrGreen I	0	0	1
56	F3.4_td	qpcR:testedat	S27a housekeeping gene	SybrGreen I	0	0	1
57	F4.1_td	qpcR:testedat	S27a housekeeping gene	SybrGreen I	0	0	1
58	F4.2_td	qpcR:testedat	S27a housekeeping gene	SybrGreen I	0	0	1
59	F4.3_td	qpcR:testedat	S27a housekeeping gene	SybrGreen I	0	0	1
60	F4.4_td	qpcR:testedat	S27a housekeeping gene	SybrGreen I	0	0	1
61	F5.1_td	qpcR:testedat	S27a housekeeping gene	SybrGreen I	0	0	1
62	F5.2_td	qpcR:testedat	S27a housekeeping gene	SybrGreen I	0	0	1
63	F5.3_td	qpcR:testedat	S27a housekeeping gene	SybrGreen I	0	0	1
64	F5.4_td	qpcR:testedat	S27a housekeeping gene	SybrGreen I	0	0	1
65	F6.1_td	qpcR:testedat	S27a housekeeping gene	SybrGreen I	0	0	1
66	F6.2_td	qpcR:testedat	S27a housekeeping gene	SybrGreen I	0	0	1
67	F6.3_td	qpcR:testedat	S27a housekeeping gene	SybrGreen I	0	0	1
68	F6.4_td	qpcR:testedat	S27a housekeeping gene	SybrGreen I	0	0	1
69	P09_WGA	Evrogen lab experiment	Whole genome amplification	EvaGreen	1	1	1
70	F10_WGA	Evrogen lab experiment	Whole genome amplification	EvaGreen	1	1	1
71	F11_1ng/mkl	Evrogen lab experiment	BRCA1 gene	TaqMan (HEX/BHQ1)	1	1	1
72	F12_1ng/mkl	Evrogen lab experiment	BRCA1 gene	TaqMan (HEX/BHQ1)	1	1	1
73	G01_100ng/mkl	Evrogen lab experiment	BRCA1 gene	TaqMan (HEX/BHQ1)	1	1	1
74	G02_100ng/mkl	Evrogen lab experiment	BRCA1 gene	TaqMan (HEX/BHQ1)	1	1	1
75	G03_1ng/mkl	Evrogen lab experiment	BRCA1 gene	TaqMan (HEX/BHQ1)	0	0	1
76	G04_1ng/mkl	Evrogen lab experiment	BRCA1 gene	TaqMan (HEX/BHQ1)	0	0	1
77	G05_100ng/mkl	Evrogen lab experiment	BRCA1 gene	TaqMan (HEX/BHQ1)	0	0	1
78	G06_100ng/mkl	Evrogen lab experiment	BRCA1 gene	TaqMan (HEX/BHQ1)	0	0	1
79	G07_1ng/mkl	Evrogen lab experiment	BRCA1 gene	TaqMan (HEX/BHQ1)	1	1	1
80	G08_1ng/mkl	Evrogen lab experiment	BRCA1 gene	TaqMan (HEX/BHQ1)	1	1	1
81	G09_100ng/mkl	Evrogen lab experiment	BRCA1 gene	TaqMan (HEX/BHQ1)	1	1	1
82	G10_100ng/mkl	Evrogen lab experiment	BRCA1 gene	TaqMan (HEX/BHQ1)	1	1	1
83	G11_1ng/mkl	Evrogen lab experiment	BRCA1 gene	TaqMan (HEX/BHQ1)	0	0	1
84	G12_1ng/mkl	Evrogen lab experiment	BRCA1 gene	TaqMan (HEX/BHQ1)	0	0	1
85	H01_100ng/mkl	Evrogen lab experiment	BRCA1 gene	TaqMan (HEX/BHQ1)	0	0	1
86	H02_100ng/mkl	Evrogen lab experiment	BRCA1 gene	TaqMan (HEX/BHQ1)	0	0	1
87	s1	Evrogen lab experiment	NRAS gene	TaqMan (FAM/BHQ1)	1	1	1
88	s2	Evrogen lab experiment	NRAS gene	TaqMan (FAM/BHQ1)	1	1	1
89	s3	Evrogen lab experiment	NRAS gene	TaqMan (FAM/BHQ1)	1	1	1
90	s4	Evrogen lab experiment	NRAS gene	TaqMan (FAM/BHQ1)	1	1	1
91	s5	Evrogen lab experiment	NRAS gene	TaqMan (FAM/BHQ1)	1	1	1
92	s6	Evrogen lab experiment	NRAS gene	TaqMan (FAM/BHQ1)	1	1	1
93	s7	Evrogen lab experiment	NRAS gene	TaqMan (FAM/BHQ1)	0	0	1
94	s8	Evrogen lab experiment	NRAS gene	TaqMan (FAM/BHQ1)	1	1	1
95	NTC	Evrogen lab experiment	NRAS gene	TaqMan (FAM/BHQ1)	0	0	1
96	NTC	Evrogen lab experiment	NRAS gene	TaqMan (FAM/BHQ1)	0	0	1

5 Results for the analysis with hookreg() and hookregNL()

This section contains the results of the analysis of the amplification curve data with the **hookreg()** function and the **hookregNL()** function. As in the previous sections, all code was commented to make it reproducible. Some rows in Table 2 and Table 3 appear to be empty. This expected behavior may occur in cases where the corresponding functions were not able to calculate the coefficients due to a failed model fit or violation of the truncation criterion.

5.1 Results for the analysis of the hookreg.rdml data set with hookreg()

The following code was used to analyze the **hookreg.rdml** data set with **hookreg()** function. The **hookreg()** function fits a linear model to a region of interest. The linear model is used to decide if the amplification curve as a hook effect or hook effect-like curvature.

```

# Load PCRedux package to obtain the data and make the hookreg() function
# available.
library(PCRedux)

# `data` is a temporary data frame of the hook.rdm1 amplification curve data file.
# Apply the hookreg() function over the amplification curves and arrange the
# results in the data frame `res_hookreg`.

res_hookreg <- data.frame(sample=colnames(data)[-1],
                          t(sapply(2L:ncol(data), function(i) {
                            hookreg(x=data[, 1], y=data[, i])
                          })))

# Fetch the calculated parameters from the calculations with the hookreg()
# function as a table `res_hookreg_table`.

res_hookreg_table <- data.frame(sample=as.character(res_hookreg[["sample"]]),
                                intercept=signif(res_hookreg[["intercept"]], 2),
                                slope=signif(res_hookreg[["slope"]], 1),
                                hook.start=signif(res_hookreg[["hook.start"]], 0),
                                hook.delta=signif(res_hookreg[["hook.delta"]], 0),
                                p.value=signif(res_hookreg[["p.value"]], 4),
                                CI.low=signif(res_hookreg[["CI.low"]], 2),
                                CI.up=signif(res_hookreg[["CI.up"]], 2),
                                hook.fit=res_hookreg[["hook.fit"]],
                                hook.CI=res_hookreg[["hook.CI"]],
                                hook=res_hookreg[["hook"]]
)

```

Finally a pretty printout (Table 2) of the results from the **hookreg()** function for the **hookreg.rdm1** data set with the following code was prepared.

```

# Load the xtable to create a LaTeX table from the `res_hookreg_table`.
library(xtable)
options(xtable.comment=FALSE)
print(xtable(res_hookreg_table,
             caption = "Results from the hookreg() function for the hookreg.rdm1
             data set.",
             label='res_hookreg_table'),
      size = "\\tiny",
      include.rownames = FALSE,
      include.colnames = TRUE,
      caption.placement = "top",
      comment=FALSE,
      table.placement = "!ht", scalebox='0.65'
)

```

The results of the **hookreg()** function are fairly comprehensive. The meaning of the columns is as followed:

- *intercept*, is the intercept from the start of the potential hook to the end of the amplification curve.
- *slope* is the slope from the start of the potential hook to the end of the amplification curve. A negative slope is indicative for a hook effect.
- *hook.start* is the estimated starting cycle of the hook region.
- *hook.delta* is the number of cycles from the *hook.start* to the end of the amplification curve.

Table 2: Results from the `hookreg()` function for the `hookreg.rdml` data set.

sample	intercept	slope	hook.start	hook.delta	p.value	CI.low	CI.up	hook.fit	hook.CI	hook
A01*F1.1	1.20	-0.01	30.00	20.00	0.00	-0.01	-0.01	1.00	1.00	1.00
A02*F1.2	1.20	-0.01	30.00	20.00	0.00	-0.01	-0.01	1.00	1.00	1.00
A03*F2.1	1.20	-0.01	30.00	9.00	0.00	-0.01	-0.00	1.00	1.00	1.00
A04*F2.2	1.20	-0.01	30.00	9.00	0.00	-0.01	-0.00	1.00	1.00	1.00
A05*F3.1	1.10	-0.00	40.00	6.00	0.05	-0.01	0.00	0.00	0.00	0.00
A06*F3.2	1.10	-0.00	40.00	6.00	0.02	-0.01	0.00	0.00	0.00	0.00
A07*F4.1	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
A08*F4.2	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
A09*F5.1	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
A10*F5.2	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
A11*F6.1	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
A12*F6.2	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
B01*HP1	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
B02*HP2	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
B03*HP3	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
B04*HP4	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
B05*HP5	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
B06*HP6	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
B07*HP7	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
B08*HP8	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
B09*HP9	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
B10*HP10	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
B11*HP11	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
B12*HP12	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
C01*HP13	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
C02*HP14	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
C03*HP15	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
C04*HP16	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
C05*HP17	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
C06*HP18	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
C07*HP19	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
C08*HP20	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
C09*HP21	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
C10*HP22	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
C11*HP23	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
C12*HP24	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
D01*HP25	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
D02*HP26	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
D03*HP27	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
D04*HP28	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
D05*HP29	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
D06*HP30	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
D07*HP31	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
D08*HP32	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
D09*F1.1_td	1.00	-0.00	30.00	7.00	0.01	-0.00	0.00	0.00	0.00	0.00
D10*F1.2_td	1.10	-0.00	30.00	10.00	0.00	-0.00	-0.00	1.00	1.00	1.00
D11*F1.3_td	0.73	-0.02	10.00	30.00	0.05	-0.05	0.01	0.00	0.00	0.00
D12*F1.4_td	0.08	-0.00	3.00	40.00	0.42	-0.02	0.01	0.00	0.00	0.00
E01*F2.1_td	1.00	-0.00	30.00	7.00	0.04	-0.00	0.00	0.00	0.00	0.00
E02*F2.2_td	1.10	-0.00	40.00	5.00	0.15	-0.01	0.01	0.00	0.00	0.00
E03*F2.3_td	-0.13	0.00	20.00	20.00	0.90	-0.06	0.06	0.00	0.00	0.00
E04*F2.4_td	3.10	-0.08	30.00	8.00	0.24	-0.36	0.21	0.00	0.00	0.00
E05*F3.1_td	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
E06*F3.2_td	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
E07*F3.3_td	0.55	-0.02	10.00	30.00	0.09	-0.05	0.02	0.00	0.00	0.00
E08*F3.4_td	0.11	-0.00	10.00	30.00	0.84	-0.04	0.03	0.00	0.00	0.00
E09*F4.1_td	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
E10*F4.2_td	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
E11*F4.3_td	2.80	-0.08	30.00	10.00	0.14	-0.29	0.13	0.00	0.00	0.00
E12*F4.4_td	0.26	-0.02	6.00	40.00	0.08	-0.06	0.02	0.00	0.00	0.00
F01*F5.1_td	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
F02*F5.2_td	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
F03*F5.3_td	2.20	-0.06	30.00	10.00	0.06	-0.17	0.05	0.00	0.00	0.00
F04*F5.4_td	-0.08	0.00	20.00	20.00	0.89	-0.06	0.06	0.00	0.00	0.00
F05*F5.1_td	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
F06*F6.2_td	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
F07*F6.3_td	0.67	-0.02	20.00	20.00	0.24	-0.07	0.04	0.00	0.00	0.00
F08*F6.4_td	0.09	-0.00	4.00	40.00	0.73	-0.03	0.02	0.00	0.00	0.00
F09*WGA	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
F10*WGA	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
F11*1ng/mkl	2.40	-0.04	40.00	20.00	0.00	-0.06	-0.02	1.00	1.00	1.00
F12*1ng/mkl	2.30	-0.04	40.00	20.00	0.00	-0.07	-0.02	1.00	1.00	1.00
G01*100 ng/mkl	1.60	-0.03	30.00	20.00	0.00	-0.04	-0.01	1.00	1.00	1.00
G02*100 ng/mkl	1.70	-0.03	30.00	20.00	0.00	-0.05	-0.02	1.00	1.00	1.00
G03*1ng/mkl	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
G04*1ng/mkl	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
G05*100 ng/mkl	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
G06*100 ng/mkl	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
G07*1ng/mkl	3.00	-0.05	40.00	10.00	0.00	-0.06	-0.04	1.00	1.00	1.00
G08*1ng/mkl	3.00	-0.05	40.00	10.00	0.00	-0.06	-0.04	1.00	1.00	1.00
G09*100 ng/mkl	2.50	-0.04	30.00	20.00	0.00	-0.05	-0.04	1.00	1.00	1.00
G10*100 ng/mkl	2.60	-0.05	30.00	20.00	0.00	-0.05	-0.04	1.00	1.00	1.00
G11*1ng/mkl	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
G12*1ng/mkl	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
H01*100 ng/mkl	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
H02*100 ng/mkl	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
H03*s1	2.40	-0.05	30.00	20.00	0.00	-0.07	-0.03	1.00	1.00	1.00
H04*s2	2.20	-0.05	30.00	20.00	0.00	-0.07	-0.02	1.00	1.00	1.00
H05*s3	4.10	-0.08	40.00	10.00	0.00	-0.13	-0.03	1.00	1.00	1.00
H06*s4	4.60	-0.09	40.00	10.00	0.00	-0.15	-0.04	1.00	1.00	1.00
H07*s5	1.60	-0.02	30.00	20.00	0.00	-0.04	-0.01	1.00	1.00	1.00
H08*s6	1.80	-0.02	40.00	9.00	0.00	-0.03	-0.01	1.00	1.00	1.00
H09*s7	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
H10*s8	1.50	-0.01	8.00	8.00	0.00	-0.02	-0.00	1.00	1.00	1.00
H11*NTC	0.92	-0.03	10.00	30.00	0.00	-0.05	-0.01	1.00	1.00	1.00
H12*NTC	0.50	-0.01	10.00	30.00	0.19	-0.05	0.02	0.00	0.00	0.00

- *p.value* describes the significant relationship between the variables in the linear regression model.
- *CI.low* and *CI.up* are the confidence interval (low and up) for the slope parameters in the fitted linear model.
- *hook.fit* is a logical parameter indicating if the fit is significant at a default threshold of 0.005.
- *hook.CI* is a logical parameter indicating if the slope of fitted linear model is within the confidence interval (0.995).
- *hook* is a logical parameter, which combines the significance test and confidence interval test (negative slope).

5.2 Results for the analysis of the `hookreg.rdml` data set with `hookregNL()`

The following code was used to analyze the `hookreg.rdml` data set with `hookregNL()` function. The procedure is similar to the analysis with the `hookreg()` function.

The `hookreg()` function fits a six parameter sigmoidal model to amplification curve. The non-linear model

$$f(x) = c + k \cdot x + \frac{d - c}{(1 + \exp(b(\log(x) - \log(e))))^f}$$

is used to decide, based on the k parameter, if the amplification curve as a hook effect or hook effect-like curvature.

```
# Note that the PCRedux package needs to be loaded (see above).
# Load the qpcR package to prevent messages during the start.
suppressMessages(library(qpcR))

# `data` is a temporary data frame of the hook.rdml amplification curve data file.
# Apply the hookregNL() function over the amplification curves and arrange the
# results in the data frame `res_hookregNL`.
# Not that `suppressMessages()` to prevent warning messages from the qpcR package.

res_hookregNL <- data.frame(sample=colnames(data)[-1],
                             t(suppressMessages(sapply(2L:ncol(data), function(i) {
                               hookregNL(x=data[, 1], y=data[, i])
                             }))))

res_hookregNL_table <- data.frame(sample=as.character(res_hookregNL[["sample"]]),
                                   slope=signif(as.numeric(res_hookregNL[["slope"]]), 1),
                                   CI.low=signif(as.numeric(res_hookregNL[["CI.low"]]), 2),
                                   CI.up=signif(as.numeric(res_hookregNL[["CI.up"]]), 2),
                                   hook.CI=unlist(res_hookregNL[["hook"]])
)
```

Finally we prepare a pretty printout (Table 3) of the results from the **hookregNL()** function for the **hookreg.rdml** data set with the following code with the code shown next.

The results of the **hookregNL()** function are less comprehensive then from the **hookreg()** function . The meaning of the columns is as followed:

- *slope* is the slope from the start of the potential hook to the end of the amplification curve that was fitted by a six parameter model. A negative slope is indicative for a hook effect.
- *CI.low* and *CI.up* is the confidence interval (low and up) for the slope parameters in the fitted linear model.
- *hook* is a logical parameter, which combines the significance test and confidence interval test (negative slope).

```
library(xtable)
options(xtable.comment=FALSE)

print(xtable(res_hookregNL_table,
             caption = "Results from the hookregNL() function for the
hookreg.rdml data set.",
             label='res_hookregNL_table'),
      size = "\\tiny",
      include.rownames = FALSE,
      include.colnames = TRUE,
      caption.placement = "top",
      comment=FALSE,
      table.placement = "!ht", scalebox='0.65'
)
```


Table 3: Results from the hookregNL() function for the hookreg.rdml data set.

sample	slope	CI_low	CI_up	hook.CI
A01*F1.1	-0.10	-0.16	-0.12	1.00
A02*F1.2	-0.20	-0.20	-0.15	1.00
A03*F2.1	-0.09	-0.13	-0.06	1.00
A04*F2.2	-0.09	-0.12	-0.06	1.00
A05*F3.1	-0.02	-0.05	0.00	0.00
A06*F3.2	-0.02	-0.06	0.01	0.00
A07*F4.1	0.00			0.00
A08*F4.2	0.00			0.00
A09*F5.1	0.01			0.00
A10*F5.2	0.01			0.00
A11*F6.1	0.00			0.00
A12*F6.2	0.00			0.00
B01*HP1	0.01			0.00
B02*HP2	0.08			0.00
B03*HP3	0.06			0.00
B04*HP4	0.03			0.00
B05*HP5	0.04			0.00
B06*HP6	0.02			0.00
B07*HP7	-0.10			0.00
B08*HP8	0.03			0.00
B09*HP9	0.05			0.00
B10*HP10	0.05			0.00
B11*HP11	0.06			0.00
B12*HP12	0.07			0.00
C01*HP13	0.05			0.00
C02*HP14	-0.04			0.00
C03*HP15	0.08			0.00
C04*HP16	0.09			0.00
C05*HP17	0.05			0.00
C06*HP18	0.03			0.00
C07*HP19	0.10			0.00
C08*HP20	0.02			0.00
C09*HP21	0.06			0.00
C10*HP22	0.01			0.00
C11*HP23	0.10			0.00
C12*HP24	0.06			0.00
D01*HP25	0.09			0.00
D02*HP26	0.10			0.00
D03*HP27	0.10			0.00
D04*HP28	0.10			0.00
D05*HP29	0.20			0.00
D06*HP30	0.10			0.00
D07*HP31	0.10			0.00
D08*HP32	0.04			0.00
D09*F1.1_td	0.09			0.00
D10*F1.2_td	-0.05			0.00
D11*F1.3_td				0.00
D12*F1.4_td				0.00
E01*F2.1_td	0.10	0.06	0.23	0.00
E02*F2.2_td	0.05			0.00
E03*F2.3_td	-0.00			0.00
E04*F2.4_td				0.00
E05*F3.1_td	0.10	0.06	0.21	0.00
E06*F3.2_td	0.09	0.04	0.15	0.00
E07*F3.3_td	-0.00			0.00
E08*F3.4_td	-0.00			0.00
E09*F4.1_td	0.10	0.02	0.17	0.00
E10*F4.2_td	0.08	0.02	0.13	0.00
E11*F4.3_td	-0.00			0.00
E12*F4.4_td	0.00			0.00
F01*F5.1_td	0.05	0.01	0.10	0.00
F02*F5.2_td	0.05			0.00
F03*F5.3_td	-0.01			0.00
F04*F5.4_td	-0.00			0.00
F05*F6.1_td	0.03			0.00
F06*F6.2_td	0.03			0.00
F07*F6.3_td				0.00
F08*F6.4_td	-0.04			0.00
F09*WGA	-20.00	-45.00	-8.80	1.00
F10*WGA	-20.00	-37.00	-9.30	1.00
F11*1ng/mkl	-0.40			0.00
F12*1ng/mkl	-0.40			0.00
G01*100 ng/mkl	-0.40			0.00
G02*100 ng/mkl	-0.40			0.00
G03*1ng/mkl	0.02	-0.00	0.03	0.00
G04*1ng/mkl	-0.01			0.00
G05*100 ng/mkl	0.03			0.00
G06*100 ng/mkl	0.10			0.00
G07*1ng/mkl	-1.00			0.00
G08*1ng/mkl	-1.00			0.00
G09*100 ng/mkl	-1.00			0.00
G10*100 ng/mkl	-1.00			0.00
G11*1ng/mkl	-0.03			0.00
G12*1ng/mkl	-0.02			0.00
H01*100 ng/mkl	-0.10			0.00
H02*100 ng/mkl	0.01			0.00
H03*s1	-4.00			0.00
H04*s2	-4.00			0.00
H05*s3	-5.00			0.00
H06*s4	-8.00			0.00
H07*s5	-0.80			0.00
H08*s6	-0.50	-0.93	-0.10	1.00
H09*s7	0.05	0.01	0.09	0.00
H10*s8	-0.04			0.00
H11*NTC	40.00			0.00
H12*NTC	40.00			0.00

6 Comparison of the hookreg() and hookregNL() methods

The decisions from the human classification (see Table 1) and the results from the machine decision (section 5.1 and section 5.2) were aggregated in Table 4.

Finally a pretty printout (Table 4) of the aggregated data set with the following code was prepared:

```
# A simple logic was applied to improve the classification result. In this case
# the assumption was, that an amplification curve has an hook effect or hook effect-like
# curvature, if either the hookreg() or hookregNL() function are positive.
```

```
meta_hookreg <- sapply(1:nrow(res), function(i){
  ifelse(res[i, "hookreg"] == 1 || res[i, "hookregNL"] == 1, 1, 0)
})

res_out <- data.frame(Sample=res[["Sample"]], res[["Human rater"]],
  res_hookreg[["hook"]], res_hookregNL_table[["hook.CI"]],
  meta_hookreg)

colnames(res_out) <- c("Sample",
```

```

        "Human rater",
        "hookreg",
        "hookregNL",
        "hookreg and hookregNL combined"
    )

```

```

library(xtable)
options(xtable.comment=FALSE)

print(xtable(res_out, digits=0,
             caption = "Aggregated decisions from the human classification and
the results from the machine decision of the hookreg() and hookregNL()
functions.", label='method_comparison'),
      caption.placement = "top",
      scalebox='0.65')

```

The performance of the `hookreg()` and `hookregNL()` functions was analyzed with the `performeR()` function of the `PCRedux` package (Table 5). The methods were adopted from Brenner and Gefeller (1997) and Kuhn (2008). Note that the formula for the calculations of the sensitivity, specificity, precision, Negative predictive value, fall-out, false negative rate, false discovery rate, Accuracy, F1 score, Matthews correlation coefficient and kappa by Cohen are described in the documentation of the `PCRedux` package.

```

res_performeR <- signif(t(rbind(
  hookreg=performeR(res_out[["hookreg"]], res_out[["Human rater"]]),
  hookregNL=performeR(res_out[["hookregNL"]], res_out[["Human rater"]]),
  combined_hookreg=performeR(res_out[["hookreg and hookregNL combined"]],
                             res_out[["Human rater"]])
)), 4)

colnames(res_performeR) <- c("hookreg", "hookregNL", "hookreg and hookregNL")

```

```

library(xtable)
options(xtable.comment=FALSE)

print(xtable(res_performeR, digits=4,
             caption = "Analysis of the performance of both algorithms. The
performance of the individual test and the combination of the tests is shown.
Note that the classification improved if the hookreg() and hookregNL() function
were combined by a logical statement. The measure were determined with the
\\textit{performeR()} function from the \\texttt{PCRedux} package. Sensitivity,
TPR; Specificity, SPC; Precision, PPV; Negative predictive value, NPV; Fall-out,
FPR; False negative rate, FNR; False discovery rate, FDR; Accuracy, ACC; F1
score, F1; Matthews correlation coefficient, MCC, Cohen's kappa (binary
classification), $\\kappa$, label='res_performeR'),
      size = "normalsize",
      include.rownames = TRUE,
      include.colnames = TRUE,
      caption.placement = "top",
      comment=FALSE,
      table.placement = "!ht", scalebox='0.75'
    )

```

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Table 4: Aggregated decisions from the human classification and the results from the machine decision of the hookreg() and hookregNL() functions.

Sample	Human rater	hookreg	hookregNL	hookreg and hookregNL combined
1 F1.1	1	1	1	1
2 F1.2	1	1	1	1
3 F2.1	1	1	1	1
4 F2.2	1	1	1	1
5 F3.1	0	0	0	0
6 F3.2	0	0	0	0
7 F4.1	0	0	0	0
8 F4.2	0	0	0	0
9 F5.1	0	0	0	0
10 F5.2	0	0	0	0
11 F6.1	0	0	0	0
12 F6.2	0	0	0	0
13 HP1	0	0	0	0
14 HP2	0	0	0	0
15 HP3	0	0	0	0
16 HP4	0	0	0	0
17 HP5	0	0	0	0
18 HP6	0	0	0	0
19 HP7	0	0	0	0
20 HP8	0	0	0	0
21 HP9	0	0	0	0
22 HP10	0	0	0	0
23 HP11	0	0	0	0
24 HP12	0	0	0	0
25 HP13	0	0	0	0
26 HP14	0	0	0	0
27 HP15	0	0	0	0
28 HP16	0	0	0	0
29 HP17	0	0	0	0
30 HP18	0	0	0	0
31 HP19	0	0	0	0
32 HP20	0	0	0	0
33 HP21	0	0	0	0
34 HP22	0	0	0	0
35 HP23	0	0	0	0
36 HP24	0	0	0	0
37 HP25	0	0	0	0
38 HP26	0	0	0	0
39 HP27	0	0	0	0
40 HP28	0	0	0	0
41 HP29	0	0	0	0
42 HP30	0	0	0	0
43 HP31	0	0	0	0
44 HP32	0	0	0	0
45 F1.1_td	0	0	0	0
46 F1.2_td	0	1	0	1
47 F1.3_td	0	0	0	0
48 F1.4_td	0	0	0	0
49 F2.1_td	0	0	0	0
50 F2.2_td	0	0	0	0
51 F2.3_td	0	0	0	0
52 F2.4_td	0	0	0	0
53 F3.1_td	0	0	0	0
54 F3.2_td	0	0	0	0
55 F3.3_td	0	0	0	0
56 F3.4_td	0	0	0	0
57 F4.1_td	0	0	0	0
58 F4.2_td	0	0	0	0
59 F4.3_td	0	0	0	0
60 F4.4_td	0	0	0	0
61 F5.1_td	0	0	0	0
62 F5.2_td	0	0	0	0
63 F5.3_td	0	0	0	0
64 F5.4_td	0	0	0	0
65 F6.1_td	0	0	0	0
66 F6.2_td	0	0	0	0
67 F6.3_td	0	0	0	0
68 F6.4_td	0	0	0	0
69 F09~WGA	1	0	1	1
70 F10~WGA	1	0	1	1
71 F11~1ng/mkl	1	1	0	1
72 F12~1ng/mkl	1	1	0	1
73 G01~100ng/mkl	1	1	0	1
74 G02~100ng/mkl	1	1	0	1
75 G03~1ng/mkl	0	0	0	0
76 G04~1ng/mkl	0	0	0	0
77 G05~100ng/mkl	0	0	0	0
78 G06~100ng/mkl	0	0	0	0
79 G07~1ng/mkl	1	1	0	1
80 G08~1ng/mkl	1	1	0	1
81 G09~100ng/mkl	1	1	0	1
82 G10~100ng/mkl	1	1	0	1
83 G11~1ng/mkl	0	0	0	0
84 G12~1ng/mkl	0	0	0	0
85 H01~100ng/mkl	0	0	0	0
86 H02~100ng/mkl	0	0	0	0
87 s1	1	1	0	1
88 s2	1	1	0	1

Table 5: Analysis of the performance of both algorithms. The performance of the individual test and the combination of the tests is shown. Note that the classification improved if the `hookreg()` and `hookregNL()` function were combined by a logical statement. The measure were determined with the *performeR()* function from the `PCRedux` package. Sensitivity, TPR; Specificity, SPC; Precision, PPV; Negative predictive value, NPV; Fall-out, FPR; False negative rate, FNR; False discovery rate, FDR; Accuracy, ACC; F1 score, F1; Matthews correlation coefficient, MCC, Cohen's kappa (binary classification), κ

	hookreg	hookregNL	hookreg and hookregNL
TPR	0.9048	0.3333	1.0000
SPC	0.9733	1.0000	0.9733
PPV	0.9048	1.0000	0.9130
NPV	0.9733	0.8427	1.0000
FPR	0.0267	0.0000	0.0267
FNR	0.0952	0.6667	0.0000
FDR	0.0952	0.0000	0.0870
ACC	0.9583	0.8542	0.9792
F1	0.9048	0.5000	0.9545
MCC	0.8781	0.5300	0.9427
LRp	33.9300	Inf	37.5000
kappa	0.8781	0.4386	0.9411
TP	19.0000	7.0000	21.0000
TN	73.0000	75.0000	73.0000
FP	2.0000	0.0000	2.0000
FN	2.0000	14.0000	0.0000
counts	96.0000	96.0000	96.0000