R-package "speadjust" A short introduction Version 0.1-1

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This packages is still in relatively early stages of development.

This document describes briefly how to use the R-package which implements the algorithm for "Guaranteed Conditional Performance of Control Charts via Bootstrap Methods." based on Gandy and Kvaløy [2013].

Some information on how to install the package is in Appendix A. Information on how to access help can be found in Appendix B. The package is loaded by

> library(spcadjust)

1 Some simple standard usage

1.1 CUSUM charts with normality assumption

The following is a simple application for CUSUM charts, assuming that all observations are normally distributed. Based on n past in-control observations X_{-n}, \ldots, X_{-1} , the in-control mean is estimated by $\hat{\mu} = \frac{1}{n} \sum_{i=-n}^{-1} X_i$ and the in-control variance by $\hat{\sigma}^2 = \frac{1}{n-1} \sum_{i=-n}^{-1} (X_i - \hat{\mu})^2$. Based on new observations X_1, X_2, \ldots , the CUSUM chart is then defined by

$$S_0 = 0$$
, $S_t = \max(0, \frac{S_{t-1} + X_t - \hat{\mu} - \Delta/2}{\hat{\sigma}})$.

The following defines the control chart for an example data set of past observations.

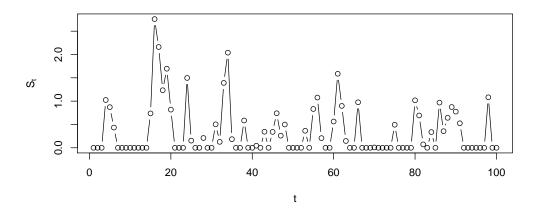
> chart <- new("SPCCUSUMNormal", Delta=1);</pre>

The following generates a data set of past observations and computes the resulting extimate for running the chart.

- > X <- rnorm(250)
 > xihat <- xiofdata(chart,X)
 > str(xihat)
 List of 3
 \$ mu: num -0 0342
- \$ mu: num -0.0342
 \$ sd: num 0.993
 \$ m : int 250

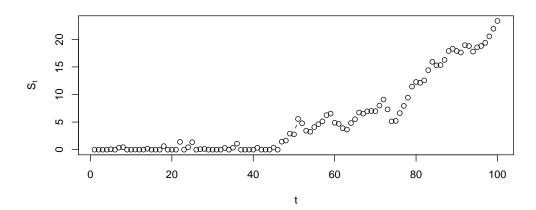
Next, we run the chart with new observations (that happen to be in-control).

> plot(runchart(chart, newdata=rnorm(100), xi=xihat), ylab=expression(S[t]), xlab="t", type="b")



In the next example, the chart is run with data that is out-of-control from time 51 onwards.

> plot(runchart(chart, newdata=rnorm(100,mean=c(rep(0,50),rep(1,50))),xi=xihat),ylab=expression(S[t]),x



The following computes confidence intervals for various properties of the chart. This is based on parametric resampling assuming normality of the observations. You should increase the number of bootstrap replications (the argument nrep) for real applications.

The first computes the threshold with a method that with roughly 90% probability results in an average run length of at least 100. The second computes a threshold such that with probability 90% this gives a false alarm probability of 5%.

Based on 50 bootstrap repetitions.

The next two examples compute confidence intervals for ARL and hitting probabilities for certain thresholds.

```
> SPCproperty(dat=X,nrep=50,
              property=new("ARLCUSUM", chart=chart, threshold=3),
              covprob = c(0.8, 0.9))
80 % CI: A threshold of 3 gives an in-control ARL of at least 83.73.
90 % CI: A threshold of 3 gives an in-control ARL of at least 68.96.
Unadjusted result: 120.6
Based on 50 bootstrap repetitions.
> SPCproperty(dat=X,nrep=50,
              property=new("hitprobCUSUM", chart=chart, threshold=5, nsteps=100),
              covprob = c(0.8, 0.9))
80 % CI: A threshold of 5 gives an in-control false alarm probability
  of at most 0.1533 within 100 steps.
90 % CI: A threshold of 5 gives an in-control false alarm probability
  of at most 0.1857 within 100 steps.
Unadjusted result: 0.0941
Based on 50 bootstrap repetitions.
```

1.2 CUSUM chart using nonparametric resampling

The following example now calibrates using nonparametric resampling (resampling the observations with replacement), which would be robust against model misspecifications.

```
> chartnp <- new("SPCCUSUMNonparCenterScale",Delta=1)
> SPCproperty(data=X,
+ nrep=100,property=new("calARLCUSUM",chart=chartnp,target=100))
90 % CI: A threshold of 3.08 gives an in-control ARL of at least 100.
Unadjusted result: 2.701
Based on 100 bootstrap repetitions.
```

1.3 Shewhart charts with normality assumptions

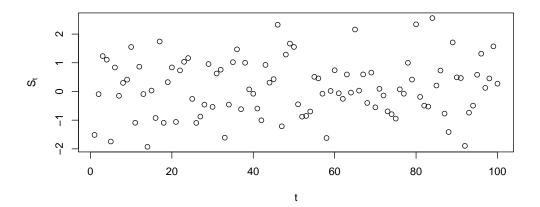
Based on n past in-control observations X_{-n}, \ldots, X_{-1} , the in-control mean is estimated by $\hat{\mu} = \frac{1}{n} \sum_{i=-n}^{-1} X_i$ and the in-control variance by $\hat{\sigma}^2 = \frac{1}{n-1} \sum_{i=-n}^{-1} (X_i - \hat{\mu})^2$. Based on new observations X_1, X_2, \ldots , the chart is then defined by

$$S_t = \frac{X_t - \hat{\mu} - \Delta/2}{\hat{\sigma}}.$$

> chartShew <- new("SPCShewNormalCenterScale")</pre>

Plotting the chart based on new data.

> plot(runchart(chartShew, newdata=rnorm(100),xi=xiofdata(chart,X)),ylab=expression(S[t]),xlab="t")



Computing various properties of the chart, adjusted for estimation error.

```
> SPCproperty(data=X,nrep=100,
              property=new("calARLShew",chart=chartShew,target=100))
90 % CI: A threshold of 2.457 gives an in-control ARL of at least 100.
Unadjusted result: 2.326
Based on 100 bootstrap repetitions.
> SPCproperty(data=X,nrep=500,
              property=new("ARLShew",chart=chartShew,threshold=4))
90 % CI: A threshold of 4 gives an in-control ARL of at least 11375.
Unadjusted result: 31574
Based on 500 bootstrap repetitions.
> SPCproperty(data=X,nrep=500,
              property=new("hitprobShew",chart=chartShew,nsteps=100,threshold=4))
90 % CI: A threshold of 4 gives an in-control false alarm probability
  of at most 0.008034 within 100 steps.
Unadjusted result: 0.003162
Based on 500 bootstrap repetitions.
> SPCproperty(data=X,nrep=500,
              property=new("calhitprobShew",chart=chartShew,target=0.01,nsteps=100))
90 % CI: A threshold of 3.967 gives an in-control false alarm
  probability of at most 0.01 within 100 steps.
Unadjusted result: 3.718
Based on 500 bootstrap repetitions.
```

2 Linear regression example

Suppose one observes past data $(Y_{-1}, X_{-1}), \ldots, (Y_{-n}, X_{-n})$, where Y_i is a response of interest and X_i is a corresponding vector of covariates. Parameter β of a linear model $EY = X\beta$, estimated via the function Im (which does least squares estimation) gives an estimator $\hat{\beta}$. The corresponding risk-adjusted CUSUM chart for new observations $(Y_1, X_1), \ldots, (Y_n, X_n)$ is

$$S_0 = 0$$
, $S_t = \max(0, S_{t-1} + Y_i - X\hat{\beta} - \frac{\Delta}{2})$,

```
Creating some past data.
> n <- 1000
> Xlinreg <- data.frame(x1= rbinom(n,1,0.4), x2= runif(n,0,1), x3= rnorm(n))</pre>
> Xlinreg$y <- 2 + Xlinreg$x1 + Xlinreg$x2 + Xlinreg$x3 + rnorm(n)</pre>
Defining a chart for a specific linear model and finding the threshold that would give an ARL of 100.
> chartlinreg <- new("SPCCUSUMlm",Delta=1,formula="y~x1+x2+x3")</pre>
> SPCproperty(data=Xlinreg,
              nrep=100,
+
               property=new("calARLCUSUM", chart=chartlinreg, target=100))
90 % CI: A threshold of 2.922 gives an in-control ARL of at least 100.
Unadjusted result: 2.699
Based on 100 bootstrap repetitions.
Same as before but for a different linear model.
> chartlinreg <- new("SPCCUSUMlm",Delta=1,formula="y~x1")</pre>
> SPCproperty(data=Xlinreg,
               nrep=100,
              property=new("calARLCUSUM",chart=chartlinreg,target=100))
90 % CI: A threshold of 5.37 gives an in-control ARL of at least 100.
Unadjusted result: 4.997
Based on 100 bootstrap repetitions.
>
```

3 Logistic regression example

where $\Delta > 0$ would be the minimal mean-shift of interest.

Suppose one observes past in-control data $(Y_{-1}, X_{-1}), \ldots, (Y_{-n}, X_{-n})$, where Y_i is a binary response variable and X_i is a corresponding vector of covariates. Suppose that in control logit($P(Y_i = 1|X_i)$) = $X_i\xi$. The function glm can be used to obtain an estimate $\hat{\xi}$.

For detecting a change to logit($P(Y_i = 1|X_i)$) = $\Delta + X_i\xi$, a CUSUM chart ratio can be defined by [Steiner et al., 2000]

$$S_t = \max(0, S_{t-1} + R_t), \quad S_0 = 0,$$

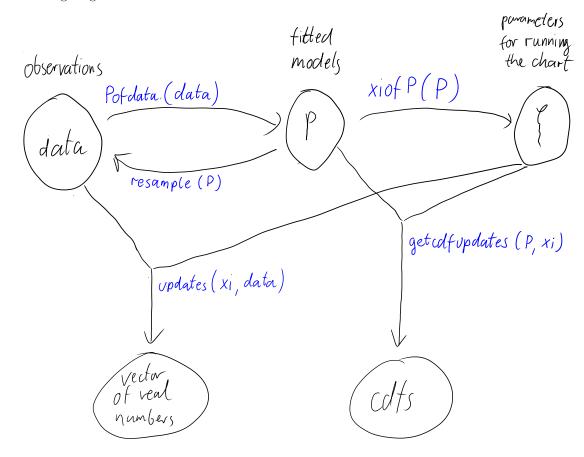
where

$$\exp(R_t) = \frac{\exp(\Delta + X_t \xi)^{Y_t} / (1 + \exp(\Delta + X_t \xi))}{\exp(X_t \xi)^{Y_t} / (1 + \exp(X_t \xi))} = \exp(Y_t \Delta) \frac{1 + \exp(X_t \xi)}{1 + \exp(\Delta + X_t \xi)}.$$

Create an example data set of past observations.

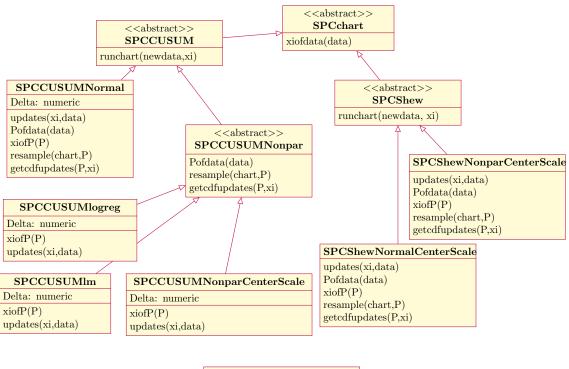
4 Methods needed for SPCproperty

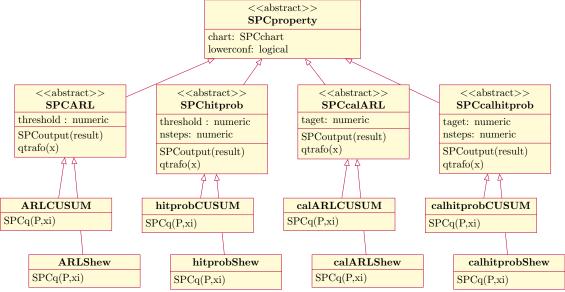
The main function SPC property needs a definition of the chart as an S4 class that has the methods marked in blue in the following diagramm.



5 Class structure

The following slightly simplified UML diagramm gives an overview over the main classes provided by the package.





6 Extensions

The framework provided in this package can be easily extended to work with different charts and/or estimation procedures. The following are some examples of extensions.

6.1 Modifying CUSUM to use robust estimation

The following defines a CUSUM chart using normality assumptions to estimate the in-control distribution using Median and MAD. Only the method Pofdata needs to be redefined, the rest is as in SPCCUSUMNormal.

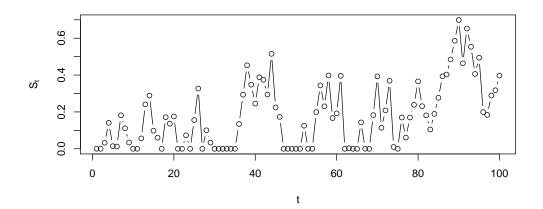
```
> setClass("SPCCUSUMNormalROBUST", contains="SPCCUSUMNormal")
> setMethod("Pofdata", "SPCCUSUMNormalROBUST",
+ function(chart,data){
+ list(mu= median(data), sd= mad(data), m=length(data))
+ })

Properties of this chart can then be computed as before:
> X <- rnorm(100)
> chartrobust <- new("SPCCUSUMNormalROBUST",Delta=1)
> SPCproperty(data=X,nrep=50,
+ property=new("calARLCUSUM",chart=chartrobust,target=100))
90 % CI: A threshold of 4.626 gives an in-control ARL of at least 100.
Unadjusted result: 2.788
Based on 50 bootstrap repetitions.
```

6.2 Parametric exponential CUSUM chart

The following defines a class that defines a CUSUM chart for exponentially distributed data with parametric resampling.

```
> setClass("SPCCUSUMExponential",contains="SPCCUSUM",representation(Delta="numeric"))
> setMethod("Pofdata", "SPCCUSUMExponential",
            function(chart,data){
                list(lambda=1/mean(data), n=length(data))
            7)
> setMethod("xiofP", "SPCCUSUMExponential",
            function(chart,P) P$lambda)
> setMethod("resample", "SPCCUSUMExponential",
            function(chart,P) rexp(P$n,rate=P$lambda))
> setMethod("getcdfupdates", "SPCCUSUMExponential",
            function(chart, P, xi) {
                 ; function(x){ if(chart@Delta<1)</pre>
                  pmax(0,1-exp(-P$lambda*(x-log(chart@Delta)))/(xi*(1-chart@Delta))))
                  pmin(1,exp(-P$lambda*(log(chart@Delta)-x)/(xi*(chart@Delta-1))))
            7)
> setMethod("updates", "SPCCUSUMExponential",
            function(chart,xi,data) log(chart@Delta)-xi*(chart@Delta-1)*data
+ )
> ExpCUSUMchart=new("SPCCUSUMExponential", Delta=1.25)
   The following greates some past observations and then runs a chart for 100 steps with new observations.
> X < - rexp(1000)
> plot(runchart(ExpCUSUMchart, newdata=rexp(100),xi=xiofdata(ExpCUSUMchart,X)),
       ylab=expression(S[t]),xlab="t",type="b")
```



The following computes various properties of the chart.

```
> SPCproperty(data=X,
              property=new("hitprobCUSUM",chart=ExpCUSUMchart,
                threshold=1,nsteps=100),covprob=c(0.5,0.9))
50 % CI: A threshold of 1 gives an in-control false alarm probability
  of at most 0.8849 within 100 steps.
90 % CI: A threshold of 1 gives an in-control false alarm probability
  of at most 0.9232 within 100 steps.
Unadjusted result: 0.8922
Based on 100 bootstrap repetitions.
> SPCproperty(data=X,
              nrep=100,
              property=new("ARLCUSUM", chart=ExpCUSUMchart,
                threshold=3), covprob=c(0.5,0.9))
50 % CI: A threshold of 3 gives an in-control ARL of at least 925.5.
90 % CI: A threshold of 3 gives an in-control ARL of at least 478.6.
Unadjusted result: 891.5
Based on 100 bootstrap repetitions.
> SPCproperty(data=X,
              nrep=100,
              property=new("calARLCUSUM", chart=ExpCUSUMchart,
                target=1000), covprob=c(0.5, 0.9))
50 % CI: A threshold of 3.292 gives an in-control ARL of at least 1000.
90 % CI: A threshold of 3.919 gives an in-control ARL of at least 1000.
Unadjusted result: 3.163
Based on 100 bootstrap repetitions.
```

References

Axel Gandy and Jan Terje Kvaløy. Guaranteed conditional performance of control charts via bootstrap methods. Scandinavian Journal of Statistics, 40:647–668, 2013. doi: 10.1002/sjos.12006. Stefan H. Steiner, Richard J. Cook, Vern T. Farewell, and Tom Treasure. Monitoring surgical performance using risk-adjusted cumulative sum charts. *Biostat*, 1(4):441–452, 2000. doi: 10.1093/biostatistics/1.4.441. URL http://biostatistics.oxfordjournals.org/cgi/content/abstract/1/4/441.

A Installation

The installation is as for most R-packages that do not reside in CRAN. The general procedure is described in the Section 6 on "Add-on packages" in the R Manual on Istallation and Administration: http://cran.r-project.org/doc/manuals/R-admin.html.

The following is merely an adaptation of those procedures to our package.

A.1 Windows

Download the package "spcadjust_0.1-1.zip". In the graphical evironment (Rgui) use the menue option : Packets ... Install packet from local zip-file.

A.2 Linux/Unix

If you do not have write access to the package repository:

- 1. Download the package "simctest_0.1-1.tar.gz" and place it into your home directory.
- 2. If needed, create a local package directory via the following commands.

```
bash
mkdir ~/Rlibrary
echo ".libPaths(\"$HOME/Rlibrary\")" >$HOME/.Rprofile
```

3. Install the package

```
R CMD INSTALL simctest_???.tar.gz where ??? needs to be replaced by 0.1-1.
```

B Accessing the help files

This document can be accessed via

> vignette("spcadjust-intro")

Documentation of the most useful command can be obtained as follows:

> ? SPCproperty

(Rudimentary) documentation of the S4-classes in this package can be obtained e.g. via

> class ? SPCCUSUMNormal