#### HARBIN

Relative quantitation data analysis tool for real-time qPCR data

### LICENSE

Harbin

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#### SYNOPSIS

Harbin is a tool for interactive evaluation of real-time qPCR data. Gene expression analysis can be performed using a relative strategy quantitation using the standard curve method normalisation with a reference gene index. Harbin also allows for the pooling of different qPCR datasets/experiments for further analysis by assigning a score to each concentration ratio and subsequently testing if the datasets are compatible before pooling expression Differential analysis between biological conditions/groups is also possible.

### CITATION

If you use Harbin in your work, please cite:

Bester, R., Pepler, P.T., Burger, J.T. and Maree, H.J. (2016) Harbin: An analysis tool for relative quantitation of real-time qPCR data and a quantile-based bootstrap test for data pooling.

### **DEPENDENCIES**

Harbin was developed for the R statistical computing environment and will run on all major platforms (Windows, Mac OS, and Linux distributions).

Harbin is dependent on base R and additional packages (psych, car, beeswarm) available from the Comprehensive R Archive Network (CRAN). Harbin can also be use independently from R (without installation of R and the additional packages) by using the shiny web application. A web browser and an Internet connection is the only requirement.

### INSTALL

For the web application:

- 1. Check to see if you have a valid Internet connection.
- 2. Open a web browser and go to the following site:

```
https://rbester.shinyapps.io/Harbin/
```

3. Start data analysis (see USAGE)

If you are familiar with R and shiny, Harbin can be run directly from R:

1. Download R or R studio from:

```
https://cran.rstudio.com/
https://www.rstudio.com/products/rstudio/download/
```

- 2. After installation load R in terminal or open the R studio app.
- 3. Use the following code in R/RStudio to check if packages are installed and install them if they are not:

```
pkg <- c("shiny", "psych", "car", "beeswarm")
new.pkg <- pkg[!(pkg %in% installed.packages())]
if (length(new.pkg)) {
  install.packages(new.pkg)
}</pre>
```

4. Download the zip Harbin directory from Github:

```
http://rbester18.github.com/harbin/
```

5. After download, UNZIP the Harbin directory and set your working path in R or RStudio to the Harbin directory:

```
setwd("Path to the unzipped harbin directory")
```

6. Load the shiny library in R/RStudio:

```
library(shiny)
```

7. Run the Harbin app:

```
runApp("Harbin_app_new_RG")
```

- 8. A second window will open with the Harbin application. Click on the "Open in Browser" button in the left upper corner to open the application in your default web browser for better visualisation.
- 9. Start data analysis (see USAGE)

# **USAGE**

The Harbin web application is organised into five different panels:

### A: Data upload

This is the default active tab when the application starts. Two options are available, either direct input of your gene of interest files and reference genes files from the Rotor-Gene Q software (version 2.3.11 and above) or manual import of Cq values from another platform.

For the Rotor-Gene option, at least one gene of interest file and one reference gene file need to be provided.

If data for one gene is split over multiple .csv files, multiple files can be selected for upload.

For each gene, standard curve data and sample data need to be available in one of the files selected for upload. Standard curve samples need to be tagged as "Standard" and samples to be included in the analysis need to be tagged as "Unknown" in the "Type" column (Automatic tagging system of the Rotor-Gene Q software). The values for the "Standard" tagged samples will be used to set the minimum value and maximum valid Cq to prevent extrapolation concentrations from the standard curve beyond the standard curve range. Please note that the Rotor-Gene Q format changed in version 2.3.11 and additional rows in the header of the file and a "Color" column was added to the .csv file.

For generation of the .csv file from the Rotor-Gene Q software please visit:

https://www.qiagen.com/za/resources/resourcedetail?id=58d4a7d9-287f-4b01-85c3-5cb83db2228b&lang=en

or see the Rotor-Gene Q manual included in the github directory at:

https://github.com/Rbester18/harbin

An example .csv file is available for download in the application or in the github directory.

After data upload, row numbers, sample names and concentration values will be checked for consistency.

For normalisation to occur, every sample name in the gene of interest file(s) need to be present in all the reference gene files and every sample need to have a concentration value in the "Rep. Calc. Conc." column. If an inconsistency is detected, a warning and/or error will be shown in the "Data upload" panel.

For manual importing of Cq values from a different qPCR platform than Rotor-Gene Q, Cq values need to be provided in a single commaseparated file (.csv). Column one should contain the names of the samples, column two the gene of interest Cq values and then column three onwards the Cq values per reference gene used. Every sample in the gene of interest column will need to have a value in each of the reference gene columns. An example file is available for download in the application or in the github directory.

After upload of files are complete, normalisation will happen in the background by dividing the gene of interest value per sample by the reference gene index (geometric mean of the reference gene concentrations for each sample). The uploaded files and the normalised values can be viewed by selecting the "Rotor-Gene data output" or "Manual import data output" panels.

#### B: Harbin intervals

After normalisation, each concentration ratio (CR) is assigned a score based on the distribution of the data. The 20th, 40th, 60th and 80th percentiles of the CRs distribution are calculated and assigned a score (1-5). A CR in the lowest quantile (0-20%) is assigned a "1", and a CR in the highest quantile (80-100%) is assigned a "5".

The normalised values and the interval scores can be downloaded at the bottom of the page.

If a previous experiment (reference database) is available and you want to add the new data to the previous experiment, the two data sets can be compared to see if the data is compatible based on the distribution functions of the two data sets. A reference database example file is available for download in the app or in the github directory. Either the Kolmogorov-Smirnov test or the Harbin test can be performed to compare data sets.

For both test, the new data is added to the reference database and the interval scores of the data in the reference database will adjust according to the new data distribution. The percentage of the elements in the reference database for which the "labels" (1-5) have changed are calculated.

The Kolmogorov-Smirnov test is a well-known test to assess the location, scale or shape of the empirical distribution functions of data sets and is the default option to compare data sets in the Harbin application. The Harbin test is proposed for a more conservative approach to avoid considering samples from two different distributions as originating from populations with the same distribution. The Harbin test is also applicable for scenarios with a larger reference database than the test data set.

Both tests will produce a p value to assess the null hypothesis that both data sets have the same distribution function. If the p value is < 0.05, the alternative hypothesis is true and the data distributions functions are different. For more details on the Harbin test, please see the end of this document.

Even though the analysis was performed using the combined data set, the new data has not been added to the database until the option to add it has been selected. After selecting this option, the new database can be downloaded with the updated interval scores. The application will also check the names of the samples present the database. If the new data set contains samples with the same names of samples in the reference database, a warning will be shown.

This panel also has a view option for the data intervals. By selecting "View intervals", plots will be displayed for the new data, unchanged database (if option to compare to database was

selected) and the new database (if option to add to reference database was selected). In these plots the different data distributions can be viewed and the influence of the distribution of each data set on the interval boundaries can be seen (indicated with dotted lines).

## C: Group selection

If applicable, the normalised data, the new reference database or a different file (formatted as reference database file) can be loaded and grouped into biological conditions/groups to perform statistical analysis. In this panel the user can select the number of groups to be compared and subsequently the same number of tables (sample name and sample value) will show up for selection of the individual samples to be classified into each group.

Parametric statistical test assumes that the variables are normally distributed and varianace homogeneous. A violation of the assumption of normality can seriously increase the chances of a Type I or II error. By transforming the data points can improve the normality and variance of variables. These transformed values can then be used in the statistical tests.

The Harbin application allows for data transformation using the natural log or log base 10.

### D: Data distribution

In this panel the basic statistics of each group selected in the previous panel will be displayed. A bar plot showing the data range and the mean of each group is also plotted, together with a box and whisker plot for better visualisation of each group's data distribution.

Two tests for normal distribution hypothesis testing are also performed. As a recommendation, the data is not normally distributed if the p value is > 0.05.

A test for homoscedasticity (Levene's test) is performed to assess the variance between groups. Variance is equal between groups if the p value is > 0.05.

Even though you've done a statistical test on a transformed variable, it is not recommended to report the basic statistics (means, standard deviation etc.) in transformed units. These statistics should be re-calculated using the untransformed data set.

# E: Statistical tests

In this panel the statistical significance testing results between the concentration ratios across biological conditions can be viewed.

If data were normally distributed the parametric test results should be used to assess the null hypothesis. For two independent sample groups, a T test is available. If equal variance was observed between groups the normal T test results can be used, if variance was not equal the Welsh T test is available. For three or more

independent sample groups a single factor analysis of variance (ANOVA) test will be performed.

For data not following a normal distribution, the non-parametric tests are available. For two independent sample groups the Wilcoxon rank sum test is available and for three or more independent groups, the Kruskal Wallis test can be used to analyse the difference between groups.

### Additional notes

Even though you've done a statistical test on a transformed variable, it is not recommended to report the basic statistics (means, standard deviation etc.) in transformed units. These statistics should be re-calculated using the untransformed data set by selecting the "do not transform option" in the "Group selection" panel. The "Data distribution" panel will refresh automatically.

Even though results from the Harbin or the Kolmogorov-Smirnov test can indicate that the pooling of qPCR data sets are possible, it is strongly recommended to only pool data sets that have been generated using the same RT-qPCR protocol.

### Details of the Harbin test

The pooling of different data sets is performed under the assumption that the samples originate from populations which can be described by the same probability distribution function.

A primary concern is therefore to determine whether these samples are compatible with each other. Suppose that  $x'=[x_1,\ldots,x_n]$  and  $y'=[y_1,\ldots,y_n]$  are representative samples from two continuous univariate populations, G and F, respectively. In many applications, it is of interest to determine whether the two population distributions are homogeneous. The hypothesis of interest is

$$H_0: G(x) = F(x), \qquad \text{for all } x, \tag{1}$$

where G(x) and F(x) are continuous univariate probability distribution functions describing the two populations. Compared to the number of parametric and non- parametric tests available for testing either equality of medians or homogeneity of variances for two groups, relatively few tests have been proposed to test for equality of the population distributions.

The more general location-scale-shape alternative hypothesis is

$$H_1: G(x) \neq F(x)$$
, for some  $x \in (-\infty, \infty)$ , (2)

The Harbin test is a quantile-based bootstrap test for hypothesis (1) against the general alternative in (2). The test works as follows:

Calculate the  $20^{\rm th}$ ,  $40^{\rm th}$ ,  $60^{\rm th}$  and  $80^{\rm th}$  percentiles of x, indicating these percentiles with  $Q_{20}$ ,  $Q_{40}$ ,  $Q_{60}$  and  $Q_{80}$ , respectively. Let  $g_i$ ,  $i=1,\ldots,n$  be a variable taking the values,

$$g_{i} = \begin{cases} 1 & \text{if } x_{i} \leq Q_{20}, \\ 2 & \text{if } Q_{20} < x_{i} \leq Q_{40}, \\ 3 & \text{if } Q_{40} < x_{i} \leq Q_{60}, \\ 4 & \text{if } Q_{60} < x_{i} \leq Q_{80}, \\ 5 & \text{if } x_{i} > Q_{80}. \end{cases}$$

$$(3)$$

Let  $h_i$  ,  $i=1,\dots,n$  (corresponding to the elements of  $\emph{x}$  ) be a variable taking the values,

$$h_{i} = \begin{cases} 1 & \text{if } x_{i} \leq Q_{20}^{*}, \\ 2 & \text{if } Q_{20}^{*} < x_{i} \leq Q_{40}^{*}, \\ 3 & \text{if } Q_{40}^{*} < x_{i} \leq Q_{60}^{*}, \\ 4 & \text{if } Q_{60}^{*} < x_{i} \leq Q_{80}^{*}, \\ 5 & \text{if } x_{i} > Q_{80}^{*}. \end{cases}$$

$$(4)$$

where  $Q_p^st$  indicates the  $p^{th}$  percentile of z. Let

$$c_{i} = \begin{cases} 0 \text{ if } g_{i} = h_{i}, \\ 1 \text{ if } g_{i} \neq h_{i}. \end{cases}$$
 (5)

The quantity  $\sum_{i=1}^n c_i$  is thus the number of elements in  $\mathcal{X}$  for which the "labels" (1-5) have changed in the combined data set, Z. The test statistic for hypothesis (1) is

$$u = \frac{1}{n} \sum_{i=1}^{n} c_i,\tag{6}$$

which is the proportion of the elements in x for which the labels have changed in the combined data set. To find the distribution of u under the null hypothesis, r=1000 bootstrap samples of size m are drawn from x. Let

$$z_0^{(j)} = \begin{bmatrix} x \\ y_0^{(j)} \end{bmatrix}, \qquad j = 1, \dots, r,$$
(7)

where  $y_0^{(j)}$  indicates the  $j^{th}$  bootstrap sample. Using x and  $z_0^{(j)}$ , the  $j^{th}$  bootstrap replication of the test statistic,  $u_0^{(j)}$ , is calculated as in (6).

The null hypothesis in (1) is rejected at a significance level of  $\alpha$  if the test statistic in (6) exceeds the  $100(1-\alpha)^{th}$  percentile of  $u_0' = \left[u_0^{(1)}, \dots, u_0^{(r)}\right]$ .

The choice of four percentiles in (3) was motivated by the application for which the test was developed. Changes in the number of percentiles in order to find the optimal number to maximise the power of the test was not studied for the purpose of this application. It is surmised that for large data sets, an increase in the number of quantiles will make the test more sensitive to detect differences between two population distributions, at the cost of a slight increase in computation time.

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