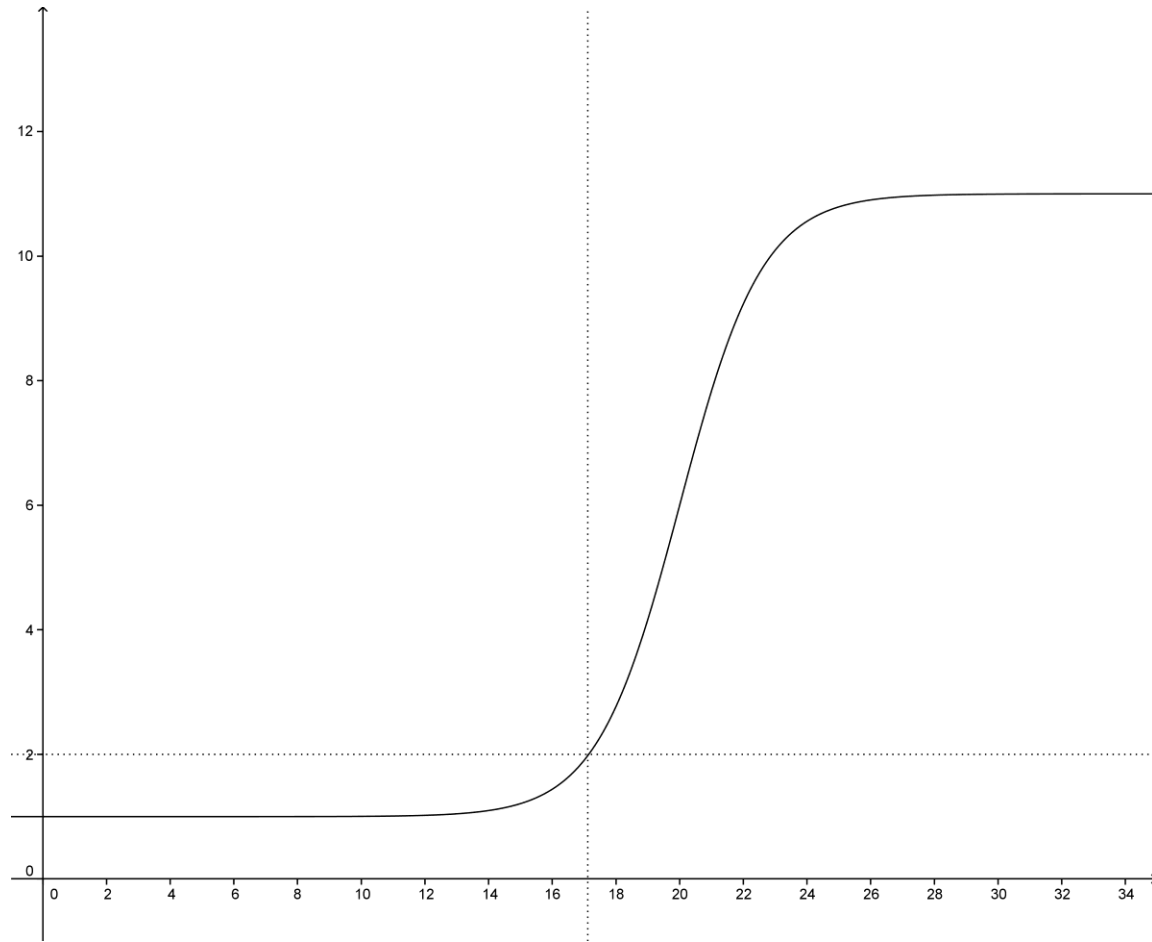


Easy analysis of qPCR data with state of the art quantification models and comprehensive quality controls using qbase^{PLUS}

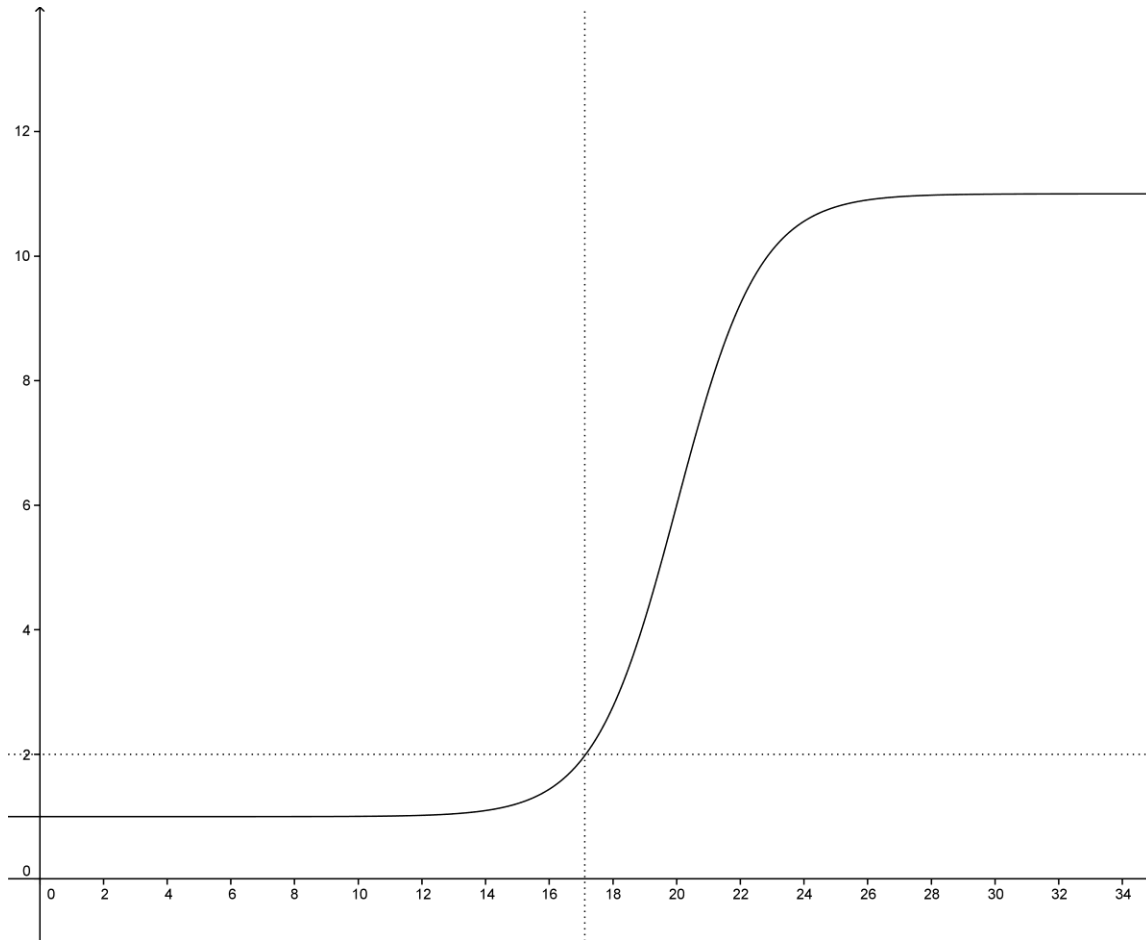
*Jan Hellemans
Ghent University
Biogazelle*



$\Delta\Delta Ct$



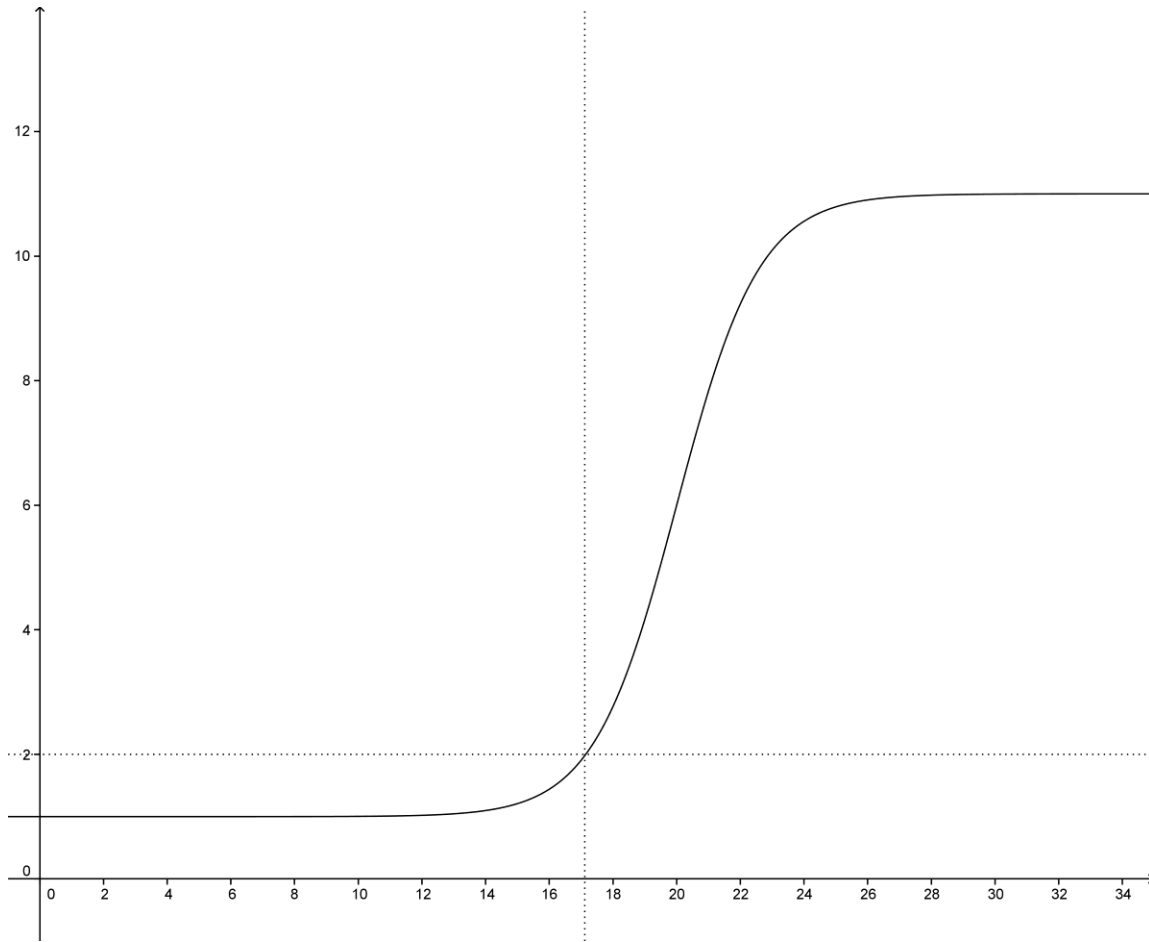
paper



$\Delta\Delta Cq$



paper



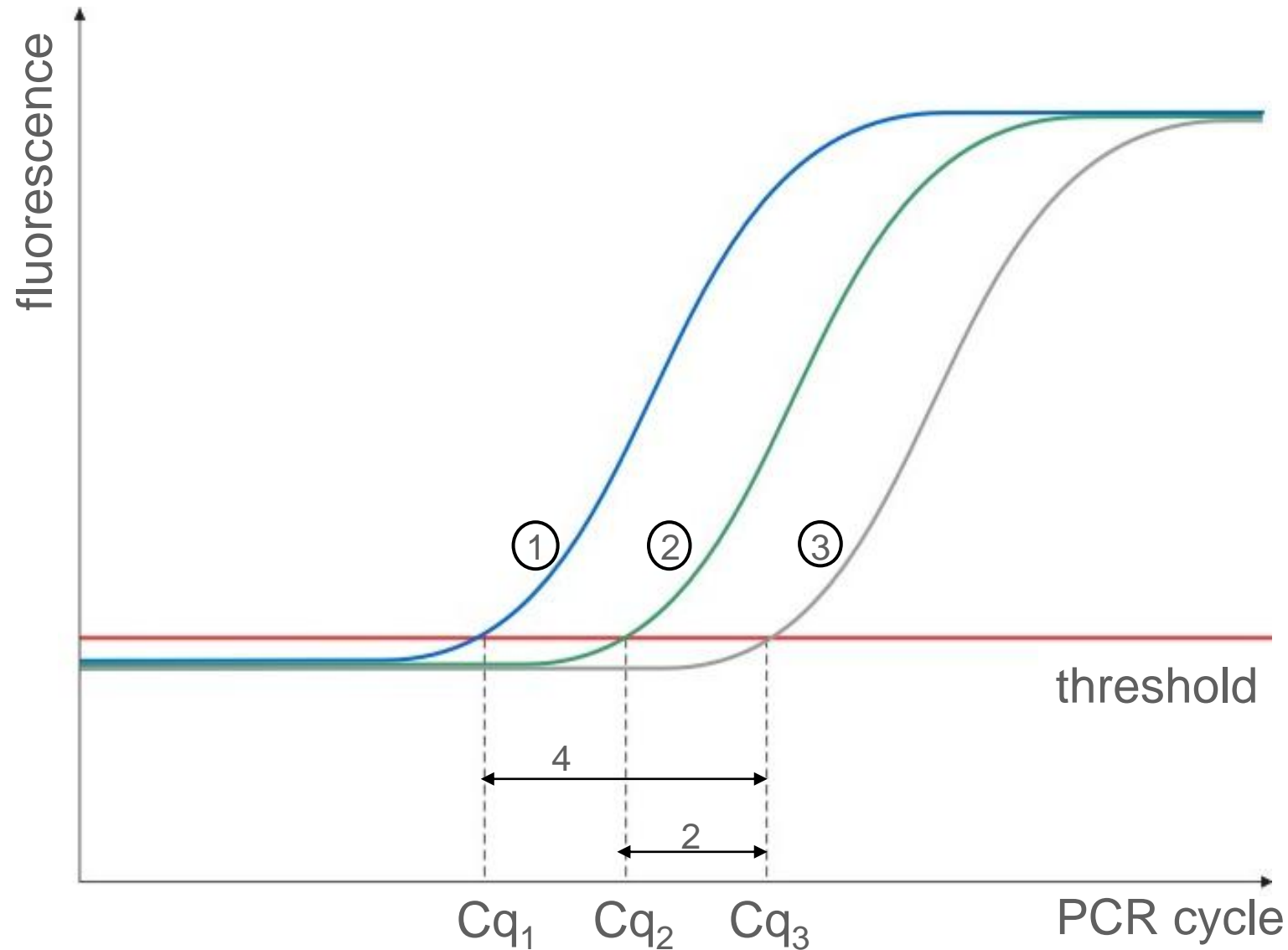
- advanced quantification models
- quality control



statistics



paper

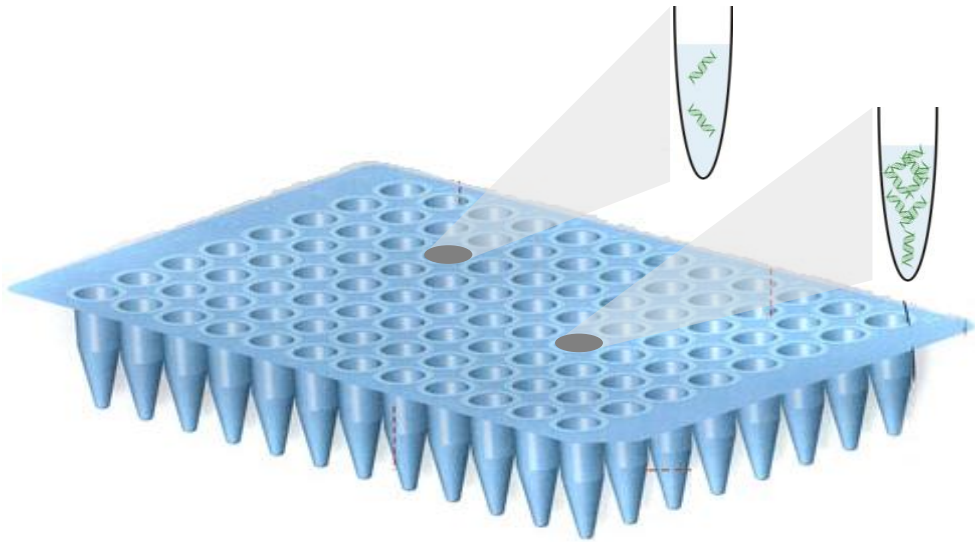


$$RQ = 2^{\Delta Cq}$$

$$RQ_{1/3} = 2^4 = 16$$

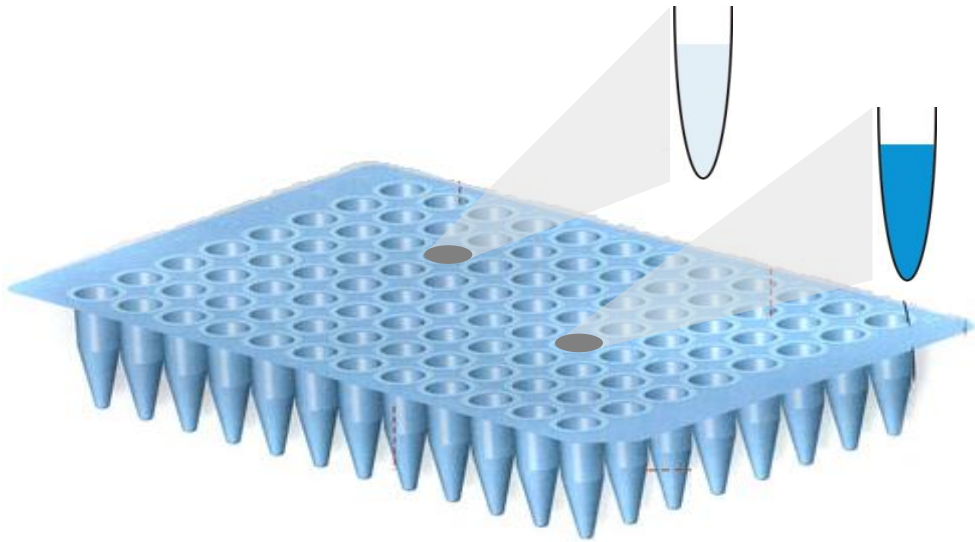
$$RQ_{2/3} = 2^2 = 4$$

$$RQ_{3/3} = 2^0 = 1$$



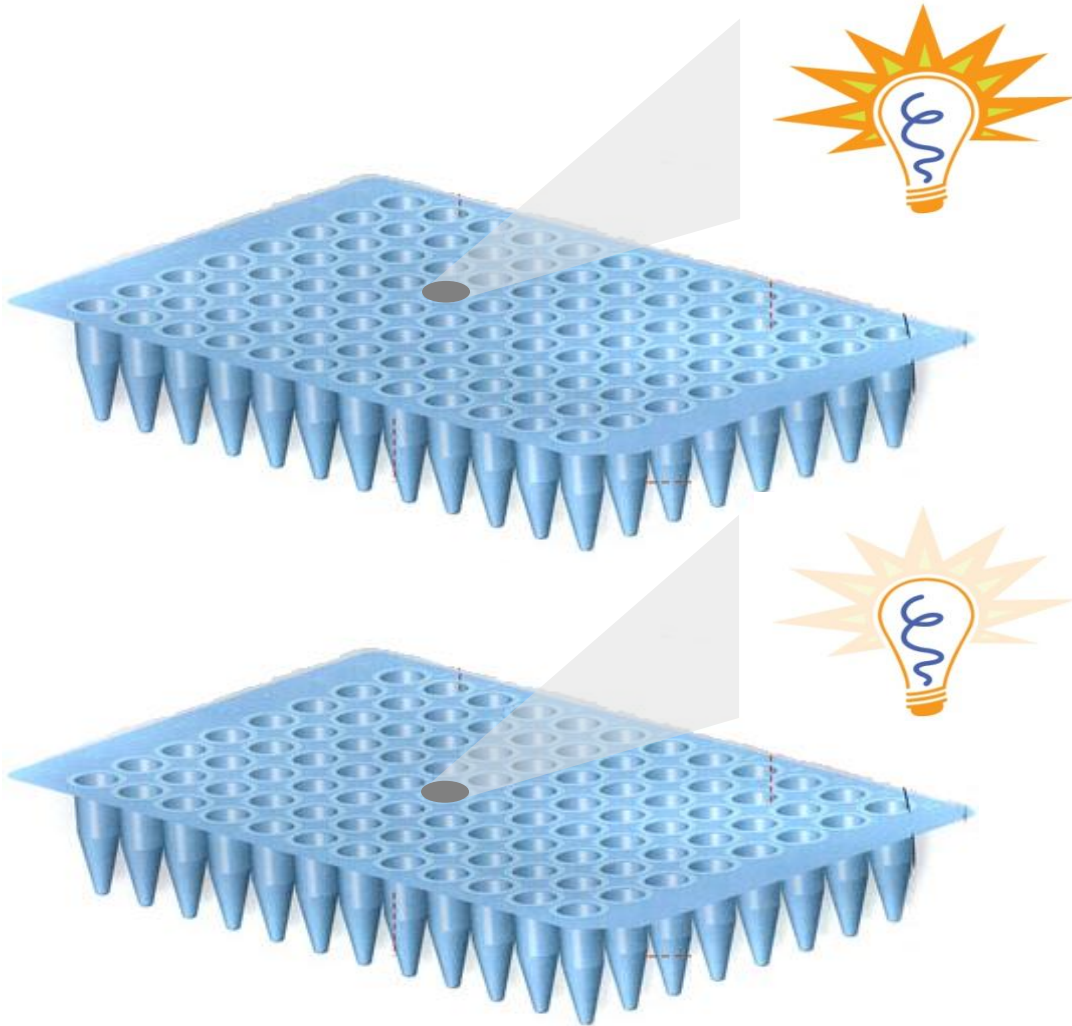
differences in RQ due to

- different gene expression level



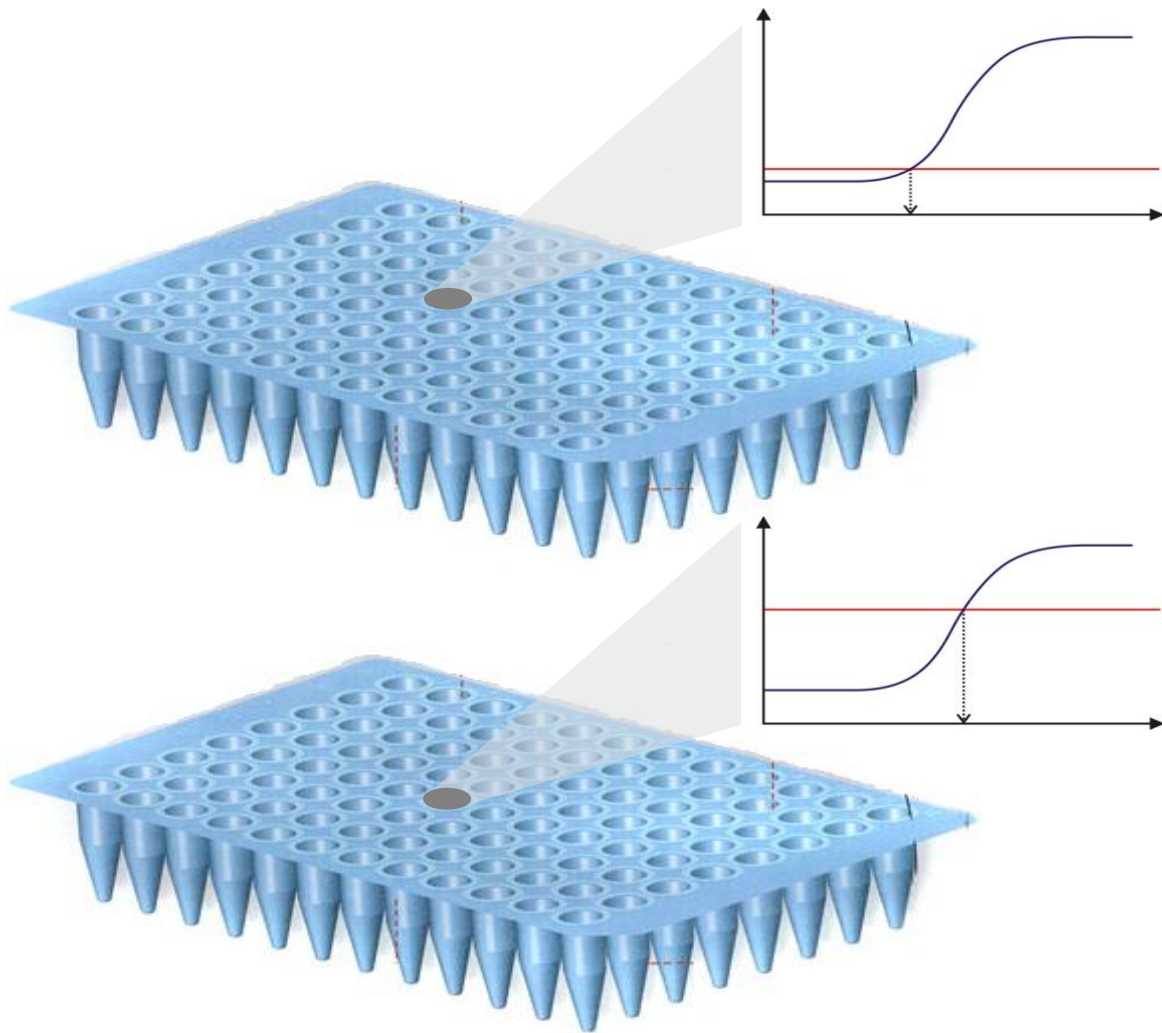
differences in RQ due to

- different gene expression level
- different total starting amount



differences in RQ due to

- different gene expression level
- different total starting amount
- run dependent differences

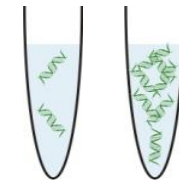
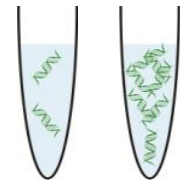
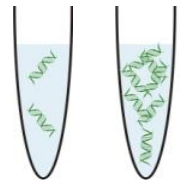


differences in RQ due to

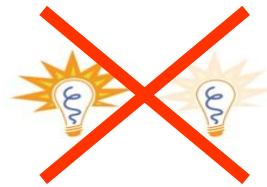
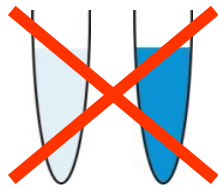
- different gene expression level
- different total starting amount
- run dependent differences

technical variation

- avoid
- minimize
- correct



Normalization



Inter-run calibration



Livak and Schmittgen (2001)

- 100% PCR efficiency
- 1 reference gene

$$NRQ = 2^{\Delta\Delta Cq}$$

Pfaffl (2001)

- adjusted PCR efficiency
- 1 reference gene

$$NRQ = \frac{E_{toi}^{\Delta Cq,toi}}{E_{ref}^{\Delta Cq,ref}}$$

Hellemans et al. (2007)

- adjusted PCR efficiency
- multiple reference genes
- qBase model

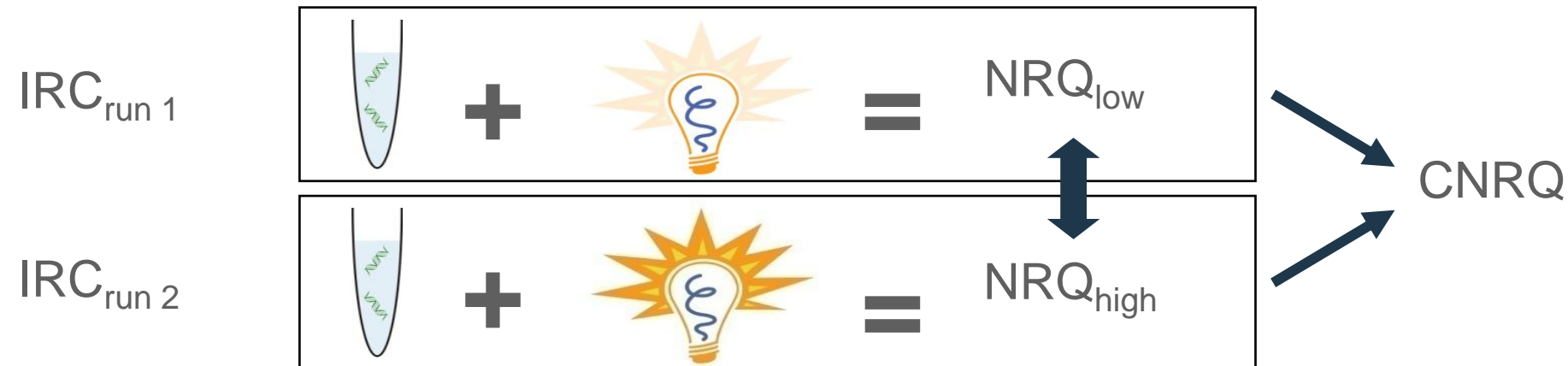
$$NRQ = \frac{E_{toi}^{\Delta Cq,toi}}{\sqrt[n]{\prod_i E_{ref_i}^{\Delta Cq,ref_i}}}$$

most flexible & most accurate

correct for inter-run variation by including IRCs

IRC:

- inter-run calibrator
- identical sample measured for the same gene in different runs



take all sources of error into account

- repeated measurements
- PCR efficiency
- normalization factor (based on multiple genes)
- calibration factor (based on multiple IRCs)

propagate the error over all calculations

true confidence in your results

geNorm normalization

- >1250 citations in PubMed
- >8000 downloads

<http://genomebiology.com/2002/3/7/research/0034.1>

Research

Accurate normalization of real-time quantitative RT-PCR data by geometric averaging of multiple internal control genes

Jo Vandesompele, Katleen De Preter, Filip Pattyn, Bruce Poppe, Nadine Van Roy, Anne De Paepe and Frank Speleman

Address: Center for Medical Genetics, Ghent University Hospital 1K5, De Pintelaan 185, B-9000 Ghent, Belgium.

$$s_{e,jl} = \sqrt{\frac{\sum_{q=1}^h (Cq_{qjl,measured} - Cq_{qjl,predicted})^2}{h-2}} \quad (\text{formula 2})$$

$$s_{x,jl} = \sqrt{\frac{1}{h-1} \sum_{q=1}^h (Q_{qjl} - \bar{Q}_{jl})^2} \quad (\text{formula 3})$$

$$SE(\text{slope}_{jl}) = \frac{s_{e,jl}}{s_{x,jl}(h-1)} \quad (\text{formula 4})$$

The base for exponential amplification E , and its standard error $SE(E)$ are calculated from these values:

$$E_{jl} = 10^{\left(\frac{1}{\text{slope}_{jl}}\right)} \quad (\text{formula 5})$$

$$SE(E_{jl}) = \frac{E_{jl} \cdot \ln(10) \cdot SE(\text{slope}_{jl})}{\text{slope}_{jl}^2} \quad (\text{formula 6})$$

Conversion of Cq values into relative quantities

Step 1
Calculation of the average Cq value for all replicates of the same gene/sample combination jk within a given run l :

$$\bar{Cq}_{jkl} = \frac{\sum_{i=1}^n Cq_{ijkl}}{n} \quad (\text{formula 7})$$

$$SE(Cq_{jkl}) = \sqrt{\frac{1}{n(n-1)} \sum_{i=1}^n (Cq_{ijkl} - \bar{Cq}_{jkl})^2} \quad (\text{formula 8})$$

Step 2
Transformation of mean Cq value into RQ using the gene specific PCR efficiency E_{jl} with minimization of the overall error:

$$Cq_{reference,jl} = \bar{Cq}_{jl} = \frac{\sum_{k=1}^s Cq_{jkl}}{s} \quad (\text{formula 9})$$

$$\Delta Cq_{jkl} = Cq_{reference,jl} - Cq_{jkl} \quad (\text{formula 10})$$

$$RQ_{jkl} = E_{jl}^{\Delta Cq_{jkl}} \quad (\text{formula 11})$$

$$SE(RQ_{jkl}) = \sqrt{RQ_{jkl}^2 \left[\left(\frac{\Delta Cq_{jkl} \cdot SD(E_{jl})}{E_{jl}} \right)^2 + (1 + E_{jl} \cdot SE(Cq_{jkl}))^2 \right]} \quad (\text{formula 12})$$

Normalization: inter-run calibration
The procedures for normalization and inter-run calibration are highly analogous and are therefore described in parallel.

Step 1
Calculation of the normalization factor NF for sample k based on the RQs of the reference genes p .

Step 1'
Calculation of the calibration factor CF for gene j in run l based on the NRQs of the IRCs m :

$$NF_k = f \sqrt{\prod_{p=1}^f RQ_{pk}} \quad (\text{formula 13})$$

$$CF_{jl} = c \sqrt{\prod_{m=1}^c NRQ_{jlm}} \quad (\text{formula 13'; for definition of NRQ, see formula 15})$$

$$SE(NF_k) = NF_k \sqrt{\sum_{p=1}^f \left(\frac{SE(RQ_{pk})}{f \cdot RQ_{pk}} \right)^2} \quad (\text{formula 14})$$

$$SE(CF_{jl}) = CF_{jl} \sqrt{\sum_{m=1}^c \left(\frac{SE(NRQ_{jlm})}{c \cdot NRQ_{jlm}} \right)^2} \quad (\text{formula 14'})$$

Step 2
Conversion of RQs into NRQs.

Step 2'
Conversion of NRQs into CNRQs:

$$NRQ_{jk} = \frac{RQ_{jk}}{NF_k} \quad (\text{formula 15})$$

$$CNRQ_{jkl} = \frac{NRQ_{jkl}}{CF_{jl}} \quad (\text{formula 15'})$$

$$SE(NRQ_{jk}) = NRQ_{jk} \sqrt{\left(\frac{SE(NF_k)}{NF_k} \right)^2 + \left(\frac{SE(RQ_{jk})}{RQ_{jk}} \right)^2} \quad (\text{formula 16})$$

$$SE(CNRQ_{jkl}) = CNRQ_{jkl} \sqrt{\left(\frac{SE(CF_{jl})}{CF_{jl}} \right)^2 + \left(\frac{SE(NRQ_{jkl})}{NRQ_{jkl}} \right)^2} \quad (\text{formula 16'})$$

Coefficient of variation of NRQs of a reference gene

Step 1
Calculation of the mean NRQ for all samples k and a given reference gene p :

$$\overline{NRQ}_p = \frac{\sum_{k=1}^s NRQ_{pk}}{s} \quad (\text{formula 17})$$

$$SE(\overline{NRQ}_p) = \sqrt{\frac{1}{s-1} \sum_{k=1}^s (NRQ_{pk} - \overline{NRQ}_p)^2} \quad (\text{formula 18})$$

Step 2
Calculation of the coefficient of variation CV of a given reference gene p across all samples k :

$$CV_p = \frac{SE(\overline{NRQ}_p)}{\overline{NRQ}_p} \quad (\text{formula 19})$$

Step 3
Calculation of the mean coefficient of variation for all reference genes:

$$\overline{CV} = \frac{\sum_{p=1}^f CV_p}{f} \quad (\text{formula 20})$$

Reference gene and IRC stability parameter M

Since normalization and inter-run calibration are highly analogous, quality evaluation using the stability parameter M is similar as well. Therefore, both methods are explained in parallel.

Step 1
Calculation of the $s \times 1$ matrix A^{gene} in which the k th element is the \log_2 transformed ratio between the relative quantities (not yet normalized) of two reference genes p and p' in sample k ; matrix A^{sample} is calculated in an analogous manner.

Step 1'
Calculation of the $g \times 1$ matrix A^{irc} in which the j th element is the \log_2 transformed ratio between the NRQs of two IRCs m and m' for the same gene j within a run l ; matrix A^{run} is calculated in an analogous manner:

$$(\forall p, p' \in [1, f], p \neq p'): A_{ppk}^{gene} = \log_2 \left(\frac{RQ_{kp}}{RQ_{kp'}} \right) \quad (\text{formula 21})$$

$$(\forall m, m' \in [1, c], m \neq m'): A_{mmj}^{irc} = \log_2 \left(\frac{NRQ_{mj}}{NRQ_{m'j}} \right) \quad (\text{formula 21'})$$

Step 2
Calculation of the mean log transformed ratio and the standard deviation V^{gene} for all samples k and a given reference gene combination (p, p') . V^{gene} is the geNorm pairwise variation V for two reference genes.

Step 2'
Calculation of the mean log transformed ratio and the standard deviation V^{irc} for all runs l and a given IRC combination (m, m') and a given gene j . V^{sample} and V^{run} are calculated similarly from A^{sample} and A^{run} , respectively:

$$A_{pp'k}^{gene} = \frac{\sum_{l=1}^r A_{pp'kl}^{gene}}{s} \quad (\text{formula 22})$$

$$A_{mm'j}^{irc} = \frac{\sum_{l=1}^r A_{mm'jl}^{irc}}{r} \quad (\text{formula 22'})$$

$$V_{pp'}^{gene} = SD(A_{pp'}^{gene}) = \sqrt{\frac{1}{s-1} \sum_{k=1}^s (A_{pp'k}^{gene} - \overline{A_{pp'}^{gene}})^2} \quad (\text{formula 23})$$

$$V_{mm'j}^{irc} = SD(A_{mm'j}^{irc}) = \sqrt{\frac{1}{r-1} \sum_{l=1}^r (A_{mm'jl}^{irc} - \overline{A_{mm'j}^{irc}})^2} \quad (\text{formula 23'})$$

Step 3
Calculation of the arithmetic mean M^{gene} of all pairwise variations V^{gene} of a given reference gene p with all other tested reference genes p' . M^{gene} represents the geNorm gene stability measure M for a particular reference gene p .

Step 3'
Calculation of the arithmetic mean M^{irc} of all pairwise variations V^{irc} of a given IRC m with all the other IRCs m' , for the same gene j . M^{sample} and M^{run} are calculated similarly from V^{sample} and V^{run} , respectively:

$$M_p^{gene} = \frac{\sum_{p'=1}^f V_{pp'}^{gene}}{f-1} \quad (\text{formula 24})$$

$$M_{mj}^{irc} = \frac{\sum_{m'=1}^c V_{mm'j}^{irc}}{c-1} \quad (\text{formula 24'})$$

Step 4
Calculation of the mean stability measure for all reference genes.

Step 4'
Calculation of the mean stability measure for all IRCs:

$$\overline{M}^{gene} = \frac{\sum_{p=1}^f M_p^{gene}}{f} \quad (\text{formula 25})$$

$$\overline{M}_j^{irc} = \frac{\sum_{m=1}^c M_{mj}^{irc}}{f} \quad (\text{formula 25'})$$

Open Access

Method

qBase relative quantification framework and software for management and automated analysis of real-time quantitative PCR data

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a simple workflow

- import Cq values from your favorite qPCR machine
- select one or more reference genes
- take a look at the results

calculations on the fly

- immediately see the impact of any change in
 - ▶ data - reference genes – settings - rescaling

based on proven geNorm and qBase quantification models

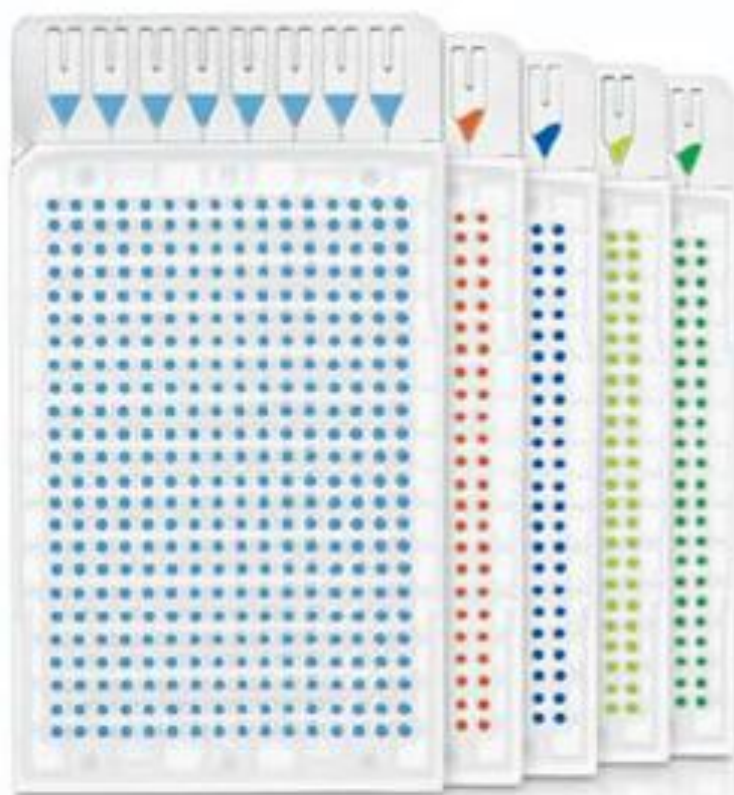
- efficiency correction
- multi-gene normalization
- inter-run calibration
- proper error propagation



00 : 00 : 00

from annotated runs
to results in 24 seconds

Analysis of more than 200 samples in triplicate for 13 genes in total took only 35 minutes in qbase^{PLUS} (including preparation of the instrument data export files). As such a job used to take 2 days of time in Excel, qBasePlus makes our lives much more comfortable.



Streamlined TaqMan analysis

- more than 10 plates
- automated analysis
- no manual interaction required

technical replicates

- $\Delta Cq < \text{threshold?}$

negative controls

- no Cq value?
- $\Delta Cq (\text{toi} - \text{ctrl}) < \text{threshold?}$

positive controls

- Cq value?

dilution series

- E close to 2 (100% efficiency)?
- linear over full range & no outliers?
- small SE(E)?

stability of reference genes

- CV of NRQs for reference genes?
- geNorm M value?

normalization factors

- no unexpected high variation?

unique set of built in quality controls

- based on user defined thresholds
- highlighting or filtering of low quality results

exclusion of bad data points

- data is not deleted
- immediate improvement of results



tutorial

get more info

- qbase^{PLUS}
- courses
- other services
- room S2 – booth 29

try it for free

- download qbase^{PLUS} from <http://www.biogazelle.com>

buy now with discount

- ask promotion code at booth

