Methods of Comparing Digital PCR Experiments

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Introduction

The outcome of digital PCR (dPCR) experiments are mean copies per partition (λ). Results are derived from the measured data, an ordered (in one or two dimensions) sequence of positive partitions. The usual analysis involves assumption the template molecules are Poisson distributed among partitions. On this premise, already proposed approaches, based on the confidence intervals (Dube et al., 2008) or uncertainty quantification (Bhat et al., 2009), allow a comparison of experiments.

GLM testing framework

Generalized Linear Models (GLM) are linear models for data, where the response variables may have non-normal distributions (as binomially distributed positive partitions in digital PCR experiments). We propose a model reflecting relationships in results of digital PCR as given by:

$$\log Y = \beta^T X \tag{1}$$

where Y are counts, X are experiments names (categorical data) and β are linear model coefficients for every experiment. Moreover, $\exp \beta = \lambda$. Estimated means copies per partitions obtained from the model are compared each other using multiple t-test (Bretz et al., 2010). The linear model above can be used only when the concentration of template molecules in samples is small (positive partitions contain very rarely more than 1 template particle). Higher concentrations basically requires binomial regression.

Multiple testing (MT) framework

The dPCR experimentes are compared pairwise using the uniformly most powerful (UMP) ratio test (Fay, 2010). Furthermore, computed p-values are adjusted using Benjamini Hochberg correction (Benjamini and Hochberg, 1995) to control family-wise error rate.

The UMP ratio test has following null-hypothesis:

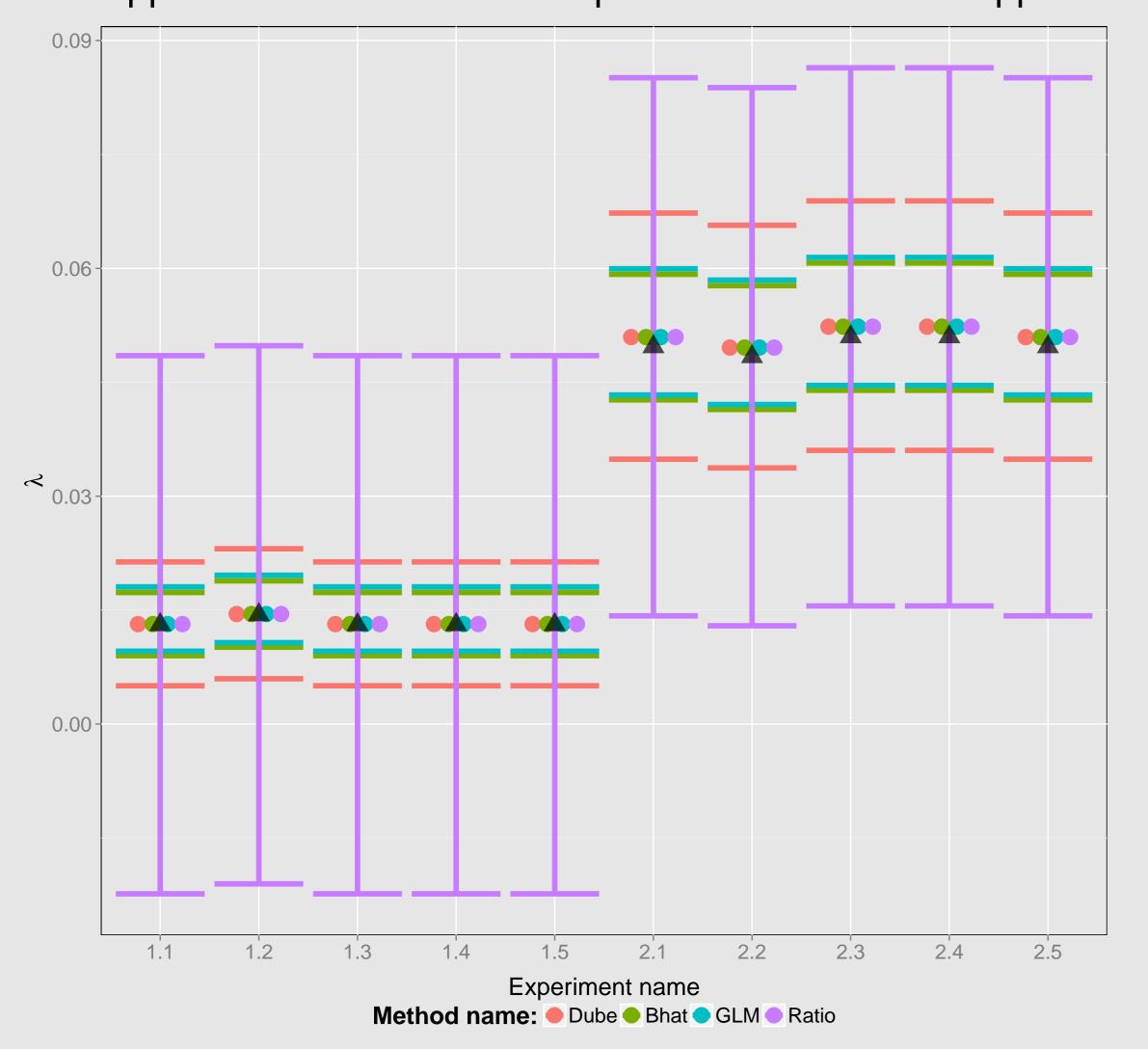
$$H_0: \frac{\lambda_1}{\lambda_2} = 1 \tag{2}$$

The Wilson's confidence intervals (Brown et al., 2001) are calculated independently for every dPCR experiment. The Dunn – Šidák correction ensures control of the family-wise error rate, because:

$$\prod_{i=1}^{T} P(\theta_{i_L} < \theta_i < \theta_{i_U}) = 1 - \alpha \tag{3}$$

Model evaluation

Two approaches above were compared in a simulation approach.



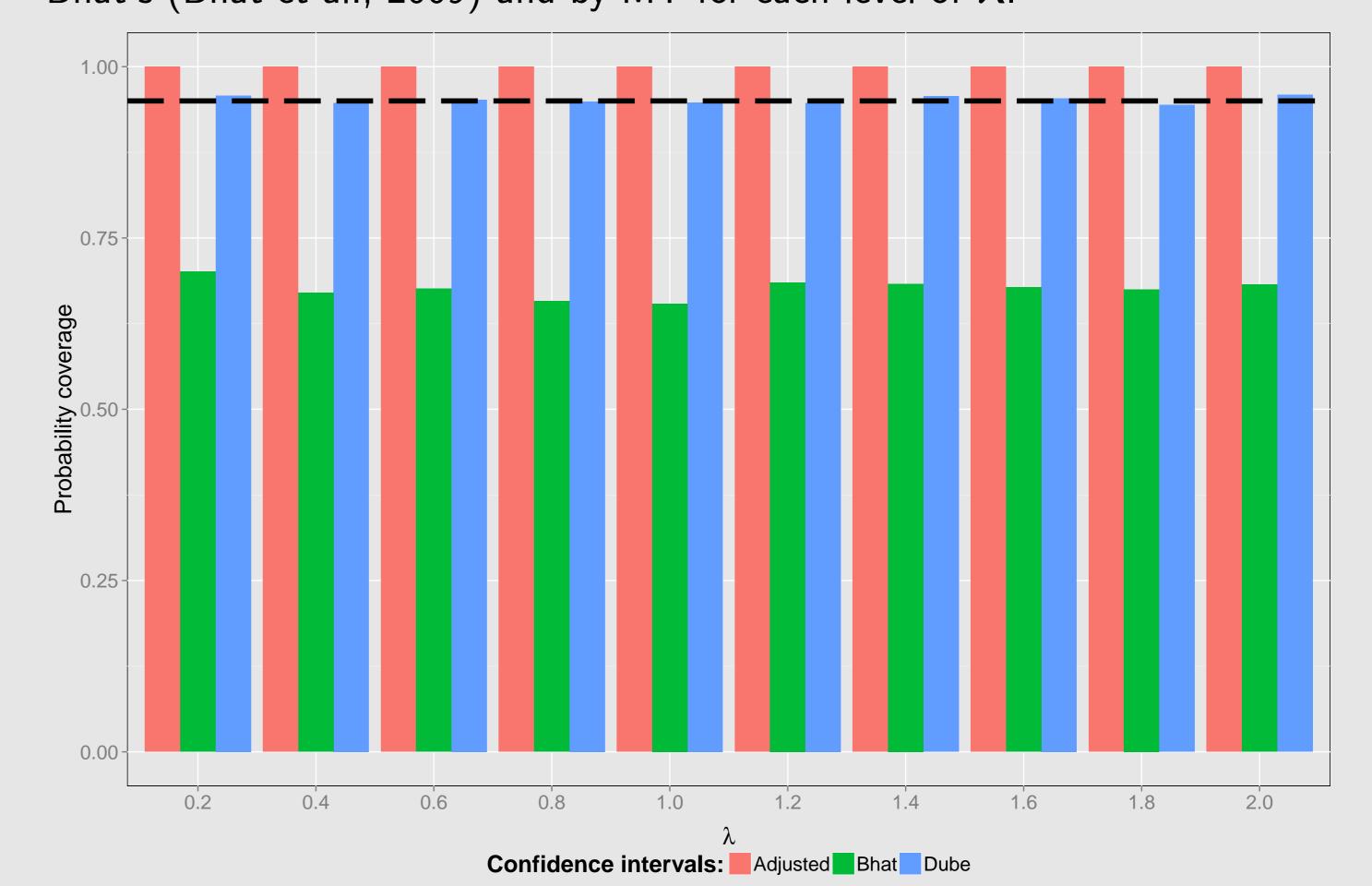
Evaluation

stuff stuff stuff

Probability coverage of confidence intervals

Covarage probability is the proportion of the time that the interval contains the true value of λ .

In the example below, we simulated 1000 droplet dPCR experiments and assed probability coverage of CI obtained by three methods: Dube's (Dube et al., 2008), Bhat's (Bhat et al., 2009) and by MT for each level of λ .



The dashed black line marks 0.95 border.

Method name	Probability coverage
Adjusted	1 00

Aajustea Bhat 0.68 Dube 0.95

On average, Dube's confidence intervals

Summary

Avaibility

dpcR web server:

dpcR R package:

http://cran.r-project.org/web/packages/dpcR/

Simulations

We performed all in-silico experiments using the accurete digital PCR data simulations (Jacobs et al., 2014).

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