An assessment of q-PCR accuracy based on RNA-seq

presented by V. Sherina

December 14, 2015

Introduction

The National Center for Biotechnology Information (NCBI) is part of the United States National Library of Medicine (NLM), a branch of the National Institutes of Health.



GEO

GEO is a public functional genomics data repository supporting MIAME-compliant data submissions. Array- and sequence-based data are accepted.

Tools are provided to help users query and download experiments and curated gene expression profiles.

Motivation

- "A comprehensive assessment of RNA-seq accuracy, reproducibility and information content by the sequencing quality control consortium." Nature Biotechnology (2014), by SEQC/MAQC-III Consortium.
- Bioconductor package with data "RNA-seq data generated from SEQC (MAQC-III) study", by Yang Liao and Wei Shi with contributions from Steve Lianoglou.

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Challenges: qPCR 96 well plate

Table: qPCR 96 well plates

Plate no	Plate start	Control gene	Plate end
1	1	42	96
2	97	98	192
3	193	194	288
4	289	311	384
5	385	412	480
6	481	516	576
7	577	579	672
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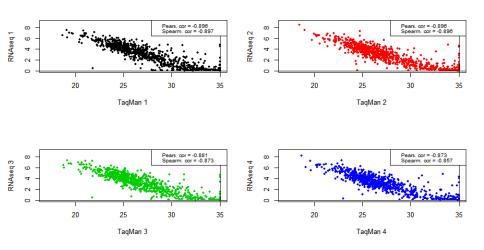
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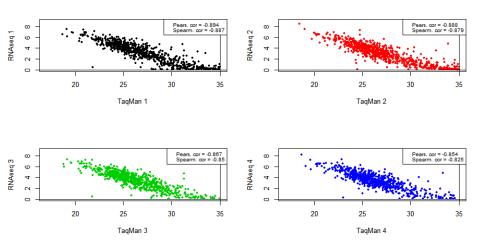
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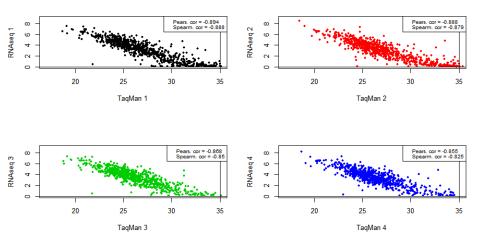


Genes absent across samples were excluded

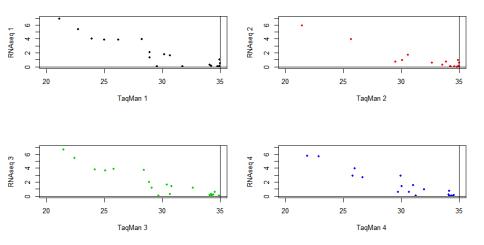


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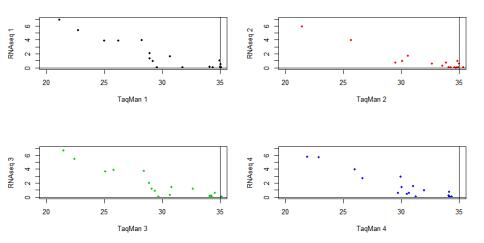
With imputed missing values



Extreme values, excluded genes



Extreme values, imputed missing data points



- CRISP3 (Cysteine-Rich Secretory Protein 3) is a Protein Coding gene. Diseases associated with CRISP3 include prostate cancer.
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- AFP Gene (sample D). This gene encodes alpha-fetoprotein, a major plasma protein produced by the yolk sac and the liver during fetal life. Alpha-fetoprotein expression in adults is often associated with hepatoma or teratoma. However, hereditary persistance of alpha-fetoprotein may also be found in individuals with no obvious pathology.

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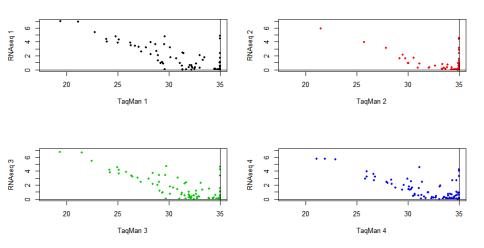
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M.N. McCall, H.R. McMurray, H. Land and A. Almudevar *On non-detects in Quantitative real-time PCR data*. Bioinformatics V. 30 no. 16, 2310-2316, 2014.



GeneCards: The Human Gene Database, http://www.genecards.org/

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