MicroRNA expression in formalin-fixed paraffin embedded tissue using real time quantitative PCR: the strengths and pitfalls

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Abstract

MicroRNAs (miRNAs) are a group of small non-coding RNAs with a huge impact in a wide range of biological processes, including cancer. The evidence collected to date demonstrates that miRNAs represent valid diagnostic, prognostic and predictive markers in cancer. The identification of these miRNA biomarkers in archived tissues has been facilitated by novel development and refinement of detection methodologies. Quantitative real-time reverse-transcription PCR (qRT-PCR) is one of the most common methods used to detect low levels of miRNAs with high sensitivity and specificity. However, several technical parameters should be identified and optimized in order to obtain meaningful and reproducible results. The purpose of this review is to describe some of these technical parameters and improve the validity and reliability of miRNA expression studies.

Keywords: microRNAs • quantitative real-time reverse transcription PCR • RNA isolation • cDNA synthesis • reference genes

Introduction

MicroRNAs (miRNAs) are a family of small, highly conserved non-coding RNAs that post-transcriptionally regulate gene expression. MiRNAs are transcribed by RNA polymerase II into long primary transcripts characterized by hairpin structures (pri-miRNAs), which are cleaved in the nucleus by Drosha, releasing a stem loop precursor miRNA (pre-miRNA). This pre-miRNA is transported to the cytoplasm where it is processed by Dicer yielding a double-stranded mature miRNA of approximately 19–22 nucleotides. After denaturation, the active single-stranded miRNA is incorporated into the RNA-induced silencing complex (RISC). The active miRNA-loaded RISC complex binds predominantly to the 3′-UTR of a target mRNA with partial miRNA sequence complementarity where the critical base pairing occurs through the 5′ seed (base pairs 2–8). Subsequently, it will either cleave the mRNA or repress its translation (Fig. 1) [1]. MiRNAs are of great interest because they can regulate approximately 30% of the human genes [2] and have a huge impact in a wide range of basic biological processes. Their regulatory functions also form the hallmarks of cancer, such as cellular differentiation, proliferation, migration and apoptosis. They can act as either oncogenes or tumour suppressor genes depending on the function of their target mRNA [3]. MicroRNAs are important epigenetic modulators because of their unique
post-transcription function, and provide more precise data than mRNA expression profiles. To date, every tumour type has shown significantly different miRNA profiles compared with normal cells from the same tissue. However, there are common miRNAs who are differentially expressed in several tumour types, such as miR-21 [4]. miR-21 is one of the first cancer miRNAs described and has been found overexpressed using qRT-PCR analysis in a variety of malignancies: breast cancer [5], colon cancer [6], glioblastoma [7], lung [8] and more. Probably this miRNA participates in fundamental signalling pathways altered in many types of cancer. Several reports highlight miR-21 to be used as a diagnostic and prognostic biomarker for cancer. In addition, beyond the promising role of miRNAs as candidate prognostic and predictive markers, there is also considerable attendance for the use of miRNAs as a novel class of therapeutic targets.

Reliable analyses of miRNA expression are a prerequisite for further research progress. Although the small size and the sequence similarity between miRNAs create challenges for their detection, recent innovative adaptations of technologies have emerged. Four main techniques have been used to quantify miRNAs including cloning, hybridization with selective probes, PCR-based detection and sequencing [9]. Cloning techniques are time-consuming and impractical for large-scale detection. The detection range for hybridization is narrow due to the fact that there is no amplification step incorporated, which limits its use in samples that yield small amounts of RNA. The most recent emerging important tool in detecting and measuring miRNA expression is next generation sequencing. This technique is mainly used as a screening tool because of the high throughput capabilities. In contrast, quantitative real-time reverse-transcription PCR (qRT-PCR) is able to detect low levels of individual miRNAs with high sensitivity and specificity on both the precursor and the mature form and is in turn less constricted by the disadvantages of cloning techniques [10]. In comparison to next generation sequencing qRT-PCR requires less highly specialized equipment and the data analysis is less complex which in turn makes it less time consuming and less costly per sample. For this reason next generation sequencing is commonly used as a screening tool, which is subsequently followed by confirmation of the finding by qRT-PCR.

To perform qRT-PCR that provides meaningful and reproducible results several parameters such as RNA extraction from formalin-fixed paraffin-embedded (FFPE) material, RNA integrity control, cDNA synthesis, amplicon detection and data normalization must be considered. The successful outcome of qRT-PCR analysis with low variability depends upon optimization (and therefore noise reduction) of all these steps. Quantitative real-time reverse transcription PCR can produce variable results even in expert laboratories, suggesting that data are only interpretable when each processing step is carried out optimally. As miRNA analysis is a fast growing research field, a previous review [11] has addressed the technical challenges associated with qRT-PCR analysis. However, our review aims to describe the remaining pitfalls associated with the specific steps in qRT-PCR analysis in more detail. Therefore, the purpose of this review is to describe the technical parameters which are relevant in miRNA expression profiling.

RNA isolation from formalin-fixed paraffin-embedded (FFPE) tissue

The most commonly available archived material used for retrospective molecular biomarker detection is FFPE tissue. Formalin fixation is a standardized method that allows preservation of cellular proteins, easy long-term storage and conservation of tissue architecture in optimal histological conditions. Unfortunately, cross linking of RNA with proteins, enzyme degradation as well as chemical degradation occurring during the fixation process reduces the yield, quality and integrity of RNA [12]. The modification rate of RNA in fixed tissue is high, showing adducts on the bases of all nucleic acids. The formation of methylol derivates, Schiff bases and stable methylene derivates at various sites negatively affects the performance of RNA samples in downstream applications [13].

Previously it has been shown that RNA samples derived from FFPE tissues can serve as effective templates for qRT-PCR [14]. However, despite optimized RNA extraction protocols, absolute
mRNA Cq values were approximately five cycles higher in FFPE tissues compared to matched fresh tissues, which indicates that only 3% of the mRNA population from FFPE tissues is accessible to cDNA synthesis and subsequent qRT-PCR analysis [15]. Even though mRNA is not stable in FFPE tissue, miRNA expression profiles do correlate well between fresh and FFPE samples [16–21]. The expression of miRNAs is preserved after routine fixation in formalin (up to 5 days) and long-term storage (10 years) [16]. MicroRNAs are less prone to degradation and modification due to their small size and close association with large protein aggregates [21]. In addition, the different miRNA types are similarly and proportionally degraded because of the uniformity in miRNA structures [14]. However, the detection of low-abundance miRNAs (not specified) might be reduced in FFPE tissue blocks stored for 7 years or longer [17]. Nevertheless, the higher extent of RNA degradation in older tissue blocks probably reduces the qRT-PCR efficiency in miRNA analysis, resulting in a higher technical variability.

There are many different isolation procedures that can be used to extract miRNAs from FFPE tissue. The different techniques rely on the subsequent steps: sample disruption and homogenization (mechanical, enzymatical, physical or chemical), lysis and purification (solution-, membrane-, magnetic bead based or by means of precipitation) [22]. These techniques have been optimized and used extensively for miRNA profiling; however, a straightforward recommendation for the optimal procedure is lacking. By comparing the performance of five commercially available kits, the RecoverAll kit (Applied Biosystems, Foster City) was identified as the preferred isolation procedure for extraction of total RNA from archived fixed tissues [18]. This protocol allows robust and reproducible isolation of high yields of total RNA, including the small miRNAs. It is recommended, in order to avoid small differences in reproducibility, to use one isolation protocol during the entire analysis.

**RNA quality control**

RNA quality control consists of measuring RNA yield, purity and integrity. Both RNA yield and purity can be measured photometrically. Pure RNA should have an OD_{260}/OD_{280} ratio higher than 2.0 which is suitable for gene expression experiments [23]. For miRNAs the same criteria can be applied. However, when using a photometrical approach, two restrictions should be made. First, DNA and RNA are detected at the same wavelengths as miRNAs and the technique is therefore non-discriminatory for these nucleic acids. Second, no conclusions about integrity can be based upon these measurements since this technique also does not discriminate for degraded forms of DNA, RNA and miRNA. To the best of our knowledge, there is no substitute technique which can overcome these imperfections. However, if the photometrical approach is used, it is important to be sure that the sample is free of DNA contamination. The most efficient strategy for this is to perform a proper DNase step within the isolation protocol. Thereafter, the absence of DNA could be evaluated performing qRT-PCR analysis of a DNA target on the total RNA [16].

RNA is a thermodynamically stable molecule, which is rapidly digested in the presence of RNase enzymes. In order to evaluate the degree of degradation, electrophoretic methods have been applied that separate the samples according to the size of the RNA fragments. Gel images show the ribosomal RNA species, and RNA is considered of high quality when the ratio of 28S:18S bands is higher than 2.0 [23]. As this showed only a weak correlation with RNA integrity, nowadays automated capillary electrophoresis is used. The amount of measured fluorescence correlates with the amount of RNA of a given size, and the subsequent software algorithm calculates the RNA integrity number (RIN) [24]. The gradual degradation of RNA is reflected by a continuous shift towards shorter fragment size and a lower RIN number. In mammalian RNA samples a minimum RIN number of 5 is suggested for reliable miRNA experiments [25]. However, the robust stability of miRNAs enables their quantification even in degraded samples of renal and prostate biopsies [26]. In the current literature there is no minimal RIN value recommended in qRT-PCR analysis. Nevertheless, for miRNA experiments using other techniques such as microarrays, a minimal RIN number of 3 has been suggested [27]. This finding might be caused by the interference of degraded RNA in the hybridization process, which in turn does not have an effect in qRT-PCR experiments. Therefore, a minimal RIN value is not always indicative in miRNA qRT-PCR measurements. Hence it proves the necessity for a RIN number to be accompanied by information relating to the mRNA and total RNA [28]. A further pitfall of assessing the RIN number is that it only determines the quality of the intact RNA.

In conclusion, currently there is no optimal technique available which is able to determine yield, purity and integrity for miRNAs. Photometrical techniques can be used to determine both yield and purity. However, it should be taken into account that these techniques measure nucleic acid and not solely miRNA concentrations. And even though techniques are available which can give an indication of the integrity of intact miRNA, no information regarding the degree of degradation is obtained. This also raises the question on how pivotal this information is, since there is no consensus on how to interpret these indirect forms of RNA measurements.

**cDNA synthesis**

Transcription into cDNA is the next step after RNA isolation. Although this process is not well understood, this pivotal step will imperatively influence the end result [29]. Underestimation of the expression or even false negative PCR reactions might be the result of an inefficient reverse transcription (RT) reaction [9]. Stahlberg et al. [30] proved that the RT yield may vary up to 100-fold depending on the choice of reverse transcriptase and that this variation is also sequence dependent.
Discrimination between the detection of the mature miRNA and its precursors is established by primer design. Synthesizing cDNA of mature miRNAs can be either initiated by universal or miRNA-specific RT (Fig. 2). The former is based upon the principle of the addition of a universal tail which is used to initiate elongation [11], while the latter uses miRNA-specific primers. MicroRNA-specific primers in turn can be either linear or contain a stem-loop [31]. Both primers consist of a 3′-end which is complementary to the miRNA. The 5′-end of the stem-loop primer comprises the stem (a double-stranded part) and loop (containing a universal binding primer binding sequence) whereas the 5′-end of the linear primer consists solely of a linear primer binding sequence. Advances of the stem-loop primers are the possible enhancement of the thermal stability of the RNA–DNA heteroduplex, due to base stacking of the stem. Second, stem-loop primers prevent the binding of double-stranded genomic DNA molecules because of spatial constraint [31]. This results in a better RT efficiency and specificity as well as an increased ability to discriminate between mature miRNAs and its precursors [31]. However, when the amount of starting material is rather small and a large number of miRNAs need to be analysed, a universal RT or the use of a multiplex RT by means of pooled stem-loop primers [32] can be beneficial. In addition, pre-amplification [33] should be performed if the total amount of cDNA input in the qRT-PCR is decreased to such a degree that it will affect the accuracy.

It is essential to carefully evaluate if replicates should be incorporated at this stage. In general, it is preferred to use replicates upstream in the hierarchy of the experiment. However, the sample type is an important determinant of the timing for incorporation of replicates [34]. The qRT-PCR variation is far greater in sampling single cells and liver tissue, compared to cell culture and blood samples. Therefore, it is recommended to use sample replicates preferentially to any other replicates when working with solid tissues. In case of blood samples, it is sufficient to use RT replicates. Introducing RT replicates results in a two-fold increment of accuracy over incorporating replicates after this stage [34]. It is also more cost-effective to introduce replicates at this stage instead of including sample replicates.

Quantitative real-time reverse transcription PCR

Quantitative real-time reverse transcription is a method for determining gene expression levels which relies on the real time detection of the amount of fluorescent reporter molecules. These reporter molecules in turn correlate directly to the number of cDNA molecules present in each cycle of amplification. Even though several fluorescent reporter molecules are used in qRT-PCR (i.e. SYBR Green I, Taqman probes, Molecular Beacons, Light upon extension and Hybprobes), only SYBR Green and Taqman probes are commonly used for performing miRNA analyses. SYBR Green I (SYBR Green) is a fluorescent DNA binding cyanide dye which is able to bind to the minor groove of double-stranded DNA [35]. Importantly, since SYBR Green binds all double-stranded DNA, discrimination for non-specific products such as primer–dimers is not possible [35]. Therefore, it is essential to analyse the dissociation curve to check for the specificity of the reaction [36]. Detection of non-specific products can be eliminated by using Taqman probes instead of SYBR Green. The probes
Reference genes

Currently, the use of endogenous control genes is the most accurate method to correct for RNA input or RT efficiency. The expression of these reference genes should remain as stable as possible in tissues or in response to experimental treatment. Many studies use well-known reference genes without proper validation of their presumed stability of expression in the tissues of interest. However, in recent years, it has also become clear that no single gene should be stable and validated. Under optimal condition the largest contributors to the run-to-run variation are the data analysis settings (base- line correction and threshold), reagents (polymerase, fluo-rophores, etc.) and optical properties of plastics. A correction can be achieved by including inter-plate calibrators. As any variation in the inter-plate calibrators will both obscure real changes and produce artificial changes in miRNA expression data, these should be stable and validated. Under optimal condition the largest contributors to the run-to-run variation are the data analysis settings. Therefore, these conditions make it possible to restrict the correction to applying a universal baseline and threshold.

Data analysis

Once the qRT-PCR experiment has been performed the data should be pre-processed. As incorrect pre-processing will give an inaccurate representation of the results, several recommendations should be taken into account. The particular steps and the order of the steps are dependent upon the design of the experiment. In this section we will discuss some of the most applied pre-processing steps, although it is important to realize that there are more steps which are not discussed here, but are just as important (Table 1).

Data analysis can be performed manually or by using dedicated software. The Cq values of the miRNA of interest and the reference gene are determined after setting the threshold. As mentioned earlier if samples are run in multiple plates it might be essential to correct for run-to-run variation, if not all miRNA assays of one sample or not all samples for one miRNA assay are on the same plate. Run-to-run variation is composed of instrument related variation (PCR block, lamp, filters, etc.), data analysis settings (baseline correction and threshold), reagents (polymerase, fluorophores, etc.) and optical properties of plastics. A correction can be achieved by including inter-plate calibrators. As any variation in the inter-plate calibrators will both obscure real changes and produce artificial changes in miRNA expression data, these should be stable and validated. Under optimal condition the largest contributors to the run-to-run variation are the data analysis settings. Therefore, these conditions make it possible to restrict the correction to applying a universal baseline and threshold.
Table 1. Overview of the steps in pre-processing qRT-PCR data

| Step no. | Description |
|----------|-------------|
| 1        | Compensate for variations between runs (inter-plate calibration) |
| 2        | Assay efficiency correction |
| 3        | Normalize with spike (efficiency variation between samples) |
| 4        | Normalize to the same amount of sample |
| 5        | Average qRT-PCR technical repeats |
| 6        | Normalize with reference genes |
| 7        | Outlier detection |
| 8        | Average other technical repeats |
| 9        | Normalize with reference samples (paired test) |
| 10       | Calculate relative quantities |
| 11       | Convert data to log scale |

If technical qRT-PCR replicates are performed, it is necessary to identify and clear the data of any outliers within the replicate set before data analysis. As a principle, outliers should only be removed if they ensue from (technical) errors. Statistical tests for outlier detection such as the Grubb’s test are available. However, the relevance of the outlier should be checked before removing it since the (repeated) application of this test can result in the loss of a substantial amount of real data. Another feasible procedure is to set a threshold that is typical for the inaccuracy of the total qRT-PCR procedure such as 0.5 cycles. Removal of replicates that deviate more than this threshold should be seriously considered. If these so-called outliers cannot be traced back to an error, qualitative agreement between analyses with and without such outliers will support the credibility of the findings.

After pre-processing the data, an appropriate method is needed to analyse and present the data. The two most commonly used methods to analyse data are absolute and relative miRNA quantification [42, 43]. In studies where the experimental objectives only require the determination of relative copy number of a target, relative quantification is commonly used instead of the standard curve method of analysis (absolute quantification). The $2^{-\Delta\Delta Cq}$ method can be used to calculate relative changes in miRNA expression determined from qRT-PCR. This form of analysis utilizes the equation:

$$\text{Relative quantity (RQ) to the calibrator} = (E + 1)^{-\Delta\Delta Cq},$$

where $E = \text{efficiency (1 = 100% efficiency)}$ and $\Delta\Delta Cq = (Cq_{\text{miRNA of interest}} - Cq_{\text{reference gene}})_{\text{sample}} - (Cq_{\text{miRNA of interest}} - Cq_{\text{reference gene}})_{\text{calibrator}}$.

Using this method, the data are presented as the fold change in miRNA expression normalized to an endogenous reference gene and relative to the calibrator. The calibrator can be any sample to which the other samples can be compared; however, the choice of the calibrator sample is usually dedicated by the type of miRNA expression experiment (i.e. tumour tissue versus normal tissue (calibrator), treated versus untreated sample (calibrator)). The $\Delta\Delta Cq$ method relies on two assumptions. First, any change in $Cq$ value for the normalizer has an equivalent effect on the $Cq$ for the miRNA of interest. This implies that the efficiency of both assays needs to be comparable or a correction factor needs to be applied in an earlier stage of the pre-processing. The second assumption is that efficiency of an assay is consistent from one run to the next [43].

Using the $\Delta\Delta Cq$ method, it is possible to calculate the RQ value of each sample relative to the calibrator, which has a clinically interpretable value. However, when comparing two groups of samples within an experiment, statistical testing is more powerful on the $\Delta Cq$ data than on the RQ data, since usually the $\Delta Cq$ data are approximately normally distributed, while the RQ data are skewed. As a result, $t$-tests can be applied to the $\Delta Cq$ data which are more powerful than the Wilcoxon–Mann–Whitney tests on the RQ data. In addition, differences between the groups can be adjusted for covariates only on $\Delta Cq$ scale, since multivariate methods assume normal (Gaussian) distribution.

Conclusions

Oscar Wilde once stated “No object is so beautiful that, under certain conditions, it will not look ugly.” The same holds true for qRT-PCR: even though the technique itself can provide meaningful and reproducible results, if every step is not carefully evaluated and optimized, it might result in the exact opposite. When designing qRT-PCR experiments for the analysis of miRNA expression in FFPE tissue, several steps should be taken into account. Some of these steps are generally applicable to all qRT-PCR experiments and others are specific for miRNA quantification.

For the isolation of miRNAs from FFPE material, several dedicated isolation procedures are available. Problems arise when the yield, purity and integrity of the isolated miRNA need to be determined. Currently, there is no single technique available that is able to determine all three parameters since techniques are non-discriminatory for miRNAs and other nucleic acids (i.e. determining yield by photometrical techniques). A pivotal step in the whole qRT-PCR procedure is the RT, since underestimation or even false negative PCR reactions are the result of an inefficient RT reaction. Therefore, it is essential to consider implementing replicates at this stage. Even though it is generally preferred to use replicates upstream in the hierarchy of the experiment, the decision relies also on cost management. Proper normalization is another critical aspect of miRNA qRT-PCR, and the suitability of reference genes should be assessed in the target tissue before start of the experiment. After pre-processing, the data should be represented using the right statistical methods.
This review intended to describe some of the essential technical parameters in microRNA qRT-PCR. Our information about the optimization of microRNA qRT-PCR experiments is coherent with the minimum information for publication of quantitative real-time PCR experiments (MIQE) [44]. This guideline constitutes a framework that promotes and improves the meaningfulness, reliability and reproducibility of qRT-PCR data.

Author contributions

J.R.D. and L.J.M.M. wrote the manuscript and were responsible for table and figure generation. I.d.K., S.T. and I.D.N. revised the manuscript critically. S.T. focussed mainly on the statistical part of the review. All authors had final approval of the submitted manuscript.

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Conflicts of interest

The authors confirm that there are no conflicts of interest.

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