

Aus der Klinik für Urologie
der Medizinischen Fakultät Charité – Universitätsmedizin Berlin und
dem Berliner Forschungsinstitut für Urologie

DISSERTATION

**Reference miRNAs for miRNAome analysis of urothelial
carcinomas**

zur Erlangung des akademischen Grades
Doctor rerum medicinalium (Dr. rer. medic.)

vorgelegt der Medizinischen Fakultät
Charité – Universitätsmedizin Berlin

von

Nadine Ratert

aus Seoul/Korea

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Nadine Ratert: Reference miRNAs for miRNAome analysis of urothelial carcinomas

Abstrakt

Hintergrund/Zielsetzung:

Die quantitative reverse Transkriptions-Polymerase-Kettenreaktion (RT-qPCR) ist eine häufig verwendete Methode zur Untersuchung von microRNA(miRNA)-Expressionen bei Tumorerkrankungen. Die relative Quantifizierung der gemessenen miRNAs, basierend auf endogenen Referenzgenen, ist dabei unerlässlich, um die Variabilität, bedingt durch die einzelnen Teilschritte der Analyse, zu kompensieren. Eine Literaturrecherche ergab, dass bis zum jetzigen Zeitpunkt keine Studien zur Ermittlung von Referenz-miRNAs für das Harnblasenkarzinom vorliegen. Das Ziel dieser Studie war es, in systematischer Weise geeignete Referenz-miRNAs für RT-qPCR basierte miRNA-Expressionsstudien des Harnblasenkarzinoms zu identifizieren.

Methodik:

Unter Zuhilfenahme eines miRNA-Microarrays wurden aus insgesamt 24 Karzinom- und Normalgewebeproben der Harnblase Kandidaten von Referenz-miRNAs anhand ihrer Invarianz in der Expressionsstabilität zwischen den Proben ermittelt. Die Validierung dieser potenziellen Referenz-miRNAs erfolgte zusammen mit den häufig in der Literatur verwendeten small RNAs, RNU6B, RNU48 und Z30, an 58 Gewebeproben mittels RT-qPCR. Die anschließende bioinformatische Analyse wurde mit den Computerprogrammen geNorm, NormFinder und BestKeeper durchgeführt.

Grundlegende Ergebnisse:

Insgesamt wurden 16 potenzielle Referenz-miRNAs auf der Grundlage der miRNA-Microarraydaten identifiziert. Nach der Validierung mittels RT-qPCR zeigten miR-101, miR-125a-5p, miR-148b, miR-151-5p, miR-181a, miR-181b, miR-29c, miR-324-3p, miR-424, miR-874, RNU6B, RNU48 und Z30 keine Unterschiede zwischen den Gewebeproben, sodass ihre Eignung als Referenzgene mit den drei Programmen ermittelt werden konnte. Daraus resultierten unterschiedliche Referenzgenkombinationen.

Schlussfolgerungen:

Die vorliegende Studie lieferte die erste systematische Analyse zur Identifizierung geeigneter Referenz-miRNAs für miRNA-Expressionsstudien des Harnblasenkarzinoms mittels RT-qPCR. Verschiedene Referenzgenkombinationen ergaben sowohl für stark- als auch für schwach-regulierte miRNAs vergleichbare Expressionsergebnisse. Besonders eindrucksvoll konnte die fehlerhafte Normalisierung mit der RNU6B belegt werden, die bisher am häufigsten in miRNA-Studien als Referenzgen zum Einsatz kam. Die Kombination aus vier (miR-101, miR-125a-5p, miR-148b und miR-151-5p) bzw. aus drei (miR-148b, miR-181b und miR-874) Referenz-miRNAs wird für die Normalisierung von Expressionsstudien beim Harnblasenkarzinom empfohlen.

Nadine Ratert: Reference miRNAs for miRNAome analysis of urothelial carcinomas

Abstract

Background/Objective:

Reverse transcription quantitative real-time PCR (RT-qPCR) is widely used in microRNA (miRNA) expression studies on cancer. To compensate for the analytical variability produced by the multiple steps of the method, relative quantification of the measured miRNAs is required, which is based on normalization to endogenous reference genes. No study has been performed so far on reference miRNAs for normalization of miRNA expression in urothelial carcinoma. The aim of this study was to identify suitable reference miRNAs for miRNA expression studies by RT-qPCR in urothelial carcinoma.

Methods:

Candidate reference miRNAs were selected from 24 urothelial carcinoma and normal bladder tissue samples by miRNA microarrays. The usefulness of these candidate reference miRNAs together with the commonly for normalization purposes used small nuclear RNAs RNU6B, RNU48, and Z30 were thereafter validated by RT-qPCR in 58 tissue samples and analyzed by the algorithms geNorm, NormFinder, and BestKeeper.

Principal Findings:

Based on the miRNA microarray data, a total of 16 miRNAs were identified as putative reference genes. After validation by RT-qPCR, miR-101, miR-125a-5p, miR-148b, miR-151-5p, miR-181a, miR-181b, miR-29c, miR-324-3p, miR-424, miR-874, RNU6B, RNU48, and Z30 were used for geNorm, NormFinder, and BestKeeper analyses that gave different combinations of recommended reference genes for normalization.

Conclusions:

The present study provided the first systematic analysis for identifying suitable reference miRNAs for miRNA expression studies of urothelial carcinoma by RT-qPCR. Different combinations of reference genes resulted in reliable expression data for both strongly and less strongly altered miRNAs. Notably, RNU6B, which is the most frequently used reference gene for miRNA studies, gave inaccurate normalization. The combination of four (miR-101, miR-125a-5p, miR-148b, and miR-151-5p) or three (miR-148b, miR-181b, and miR-874,) reference miRNAs is recommended for normalization.

Eidesstattliche Versicherung

„Ich, Nadine Ratert, versichere an Eides statt durch meine eigenhändige Unterschrift, dass ich die vorgelegte Dissertation in Form einer Publikationspromotion mit dem Thema "Reference miRNAs for miRNAome analysis of urothelial carcinomas"" mit einer Erstautorenschaft aus dem TOP-Journal PLoS One selbstständig und ohne nicht offengelegte Hilfe Dritter verfasst und keine anderen als die angegebenen Quellen und Hilfsmittel genutzt habe.

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Mein Anteil an der ausgewählten Publikation entspricht dem, der in der untenstehenden gemeinsamen Erklärung mit dem/der Betreuer/in, angegeben ist.

Die Bedeutung dieser eidesstattlichen Versicherung und die strafrechtlichen Folgen einer unwahren eidesstattlichen Versicherung (§156,161 des Strafgesetzbuches) sind mir bekannt und bewusst.“

Berlin, den 18.06.2013

Nadine Ratert

Ausführliche Anteilserklärung an der erfolgten Publikation

Publikation: Ratert N, Meyer HA, Jung M, Mollenkopf HJ, Wagner I, Miller Kurt, Kilic E, Erbersdobler A, Weikert S, Jung K. Reference miRNAs for miRNAome analysis of urothelial carcinomas. PloS One 2012;7:e39309.

Beitrag im Einzelnen:

- a)** Vorbereitung der Gewebeproben (Probenselektion anhand von Patientendaten und Überprüfung der pathohistologischen Befunde mit Hilfe des Uropathologen; RNA-Isolierung, Charakterisierung der RNA-Proben) für die miRNA-Microarray-Experimente und Chip-Hybridisierung.
- b)** Hybridisierung und Messungen am Hochleistungsscanner erfolgte in Kooperation mit dem Max-Planck-Institut für Infektionsbiologie, Berlin.
- c)** Auswertung der miRNA-Microarraydaten hinsichtlich der Expressionsintensität und Invarianz der miRNA-Expression zwischen Normal- und Tumorgewebe als Kriterien für potenzielle Referenz-miRNAs.
- d)** Validierung der ausgewählten Referenz-miRNA-Kandidaten und der zusätzlich eingeschlossenen small RNAs mittels RT-qPCR-Bestimmungen. Durchführung nach "MIQE-Guidelines".
- e)** Statistische Auswertung der Expressionsdaten und anschließende bioinformatische Analyse mit den Computerprogrammen geNorm, NormFinder, BestKeeper zur Ermittlung optimaler Kombinationen von Referenzgenen.
- f)** Interpretation der Ergebnisse, Erstellen der Abbildungen, Verfassen des Manuskripttextes und Diskussion mit dem Seniorautor (Betreuer) bis zur Fertigstellung nach Umlaufkorrektur durch alle beteiligten Autoren.

Berlin, den 18.06.2013

Nadine Ratert

Auszug aus der Journal Summary List (ISI Web of KnowledgeSM)

Die ausgewählte Publikation „Reference miRNAs for miRNAome analysis of urothelial carcinomas“ wurde erfolgreich im Juni 2012 in dem Journal PloS One veröffentlicht. Dieses nimmt den 12. Platz (Impact Factor: 4.092) von insgesamt 85 Journalen des Fachgebietes Biologie ein und befindet sich somit unter den ersten 30% der nach Impact Factor sortierten Journalen des Fachgebietes. Der Eigenfaktor des Journals PloS One liegt bei 0.50162. Somit erfüllt das Journal PloS One alle Voraussetzungen eines „Topjournals“.

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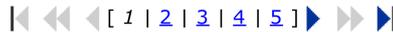
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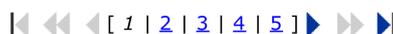


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				Total Cites	Impact Factor	5-Year Impact Factor	Immediacy Index	Articles	Cited Half-life	Eigenfactor [®] Score	Article Influence [®] Score
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<input type="checkbox"/>	2	BIOL REV	1464-7931	5678	9.067	11.790	1.830	47	8.9	0.01402	4.870
<input type="checkbox"/>	3	Q REV BIOL	0033-5770	3153	7.727	6.538	0.750	8	>10.0	0.00320	3.355
<input type="checkbox"/>	4	PHYS LIFE REV	1571-0645	549	7.208	5.241	10.917	12	2.7	0.00215	2.020
<input type="checkbox"/>	5	PHILOS T R SOC B	0962-8436	23973	6.401	7.154	2.746	319	7.1	0.07729	3.300
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<input type="checkbox"/>	14	BIOL DIRECT	1745-6150	979	4.017	3.860	0.587	63	3.0	0.00690	1.548
<input type="checkbox"/>	15	BIOL LETTERS	1744-9561	4455	3.762	4.049	0.555	247	3.7	0.02992	1.723
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Publikation

Ratert N, Meyer HA, Jung M, Mollenkopf HJ, Wagner I, Miller Kurt, Kilic E, Erbersdobler A, Weikert S, Jung K

Reference miRNAs of miRNAome Analysis of Urothelial Carcinomas

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Reference miRNAs for miRNAome Analysis of Urothelial Carcinomas

Nadine Ratert^{1,2}, Hellmuth-Alexander Meyer^{1,3}, Monika Jung¹, Hans-Joachim Mollenkopf⁴, Ina Wagner⁴, Kurt Miller¹, Ergin Kilic⁵, Andreas Erbersdobler⁶, Steffen Weikert¹, Klaus Jung^{1,2*}

1 Department of Urology, University Hospital Charité, Berlin, Germany, **2** Berlin Institute for Urologic Research, Berlin, Germany, **3** Institute of Physiology, University Hospital Charité, Berlin, Germany, **4** Max Planck Institute for Infection Biology, Berlin, Germany, **5** Institute of Pathology, University Hospital Charité, Berlin, Germany, **6** Institute of Pathology, University Rostock, Rostock, Germany

Abstract

Background/Objective: Reverse transcription quantitative real-time PCR (RT-qPCR) is widely used in microRNA (miRNA) expression studies on cancer. To compensate for the analytical variability produced by the multiple steps of the method, relative quantification of the measured miRNAs is required, which is based on normalization to endogenous reference genes. No study has been performed so far on reference miRNAs for normalization of miRNA expression in urothelial carcinoma. The aim of this study was to identify suitable reference miRNAs for miRNA expression studies by RT-qPCR in urothelial carcinoma.

Methods: Candidate reference miRNAs were selected from 24 urothelial carcinoma and normal bladder tissue samples by miRNA microarrays. The usefulness of these candidate reference miRNAs together with the commonly for normalization purposes used small nuclear RNAs RNU6B, RNU48, and Z30 were thereafter validated by RT-qPCR in 58 tissue samples and analyzed by the algorithms geNorm, NormFinder, and BestKeeper.

Principal Findings: Based on the miRNA microarray data, a total of 16 miRNAs were identified as putative reference genes. After validation by RT-qPCR, miR-101, miR-125a-5p, miR-148b, miR-151-5p, miR-181a, miR-181b, miR-29c, miR-324-3p, miR-424, miR-874, RNU6B, RNU48, and Z30 were used for geNorm, NormFinder, and BestKeeper analyses that gave different combinations of recommended reference genes for normalization.

Conclusions: The present study provided the first systematic analysis for identifying suitable reference miRNAs for miRNA expression studies of urothelial carcinoma by RT-qPCR. Different combinations of reference genes resulted in reliable expression data for both strongly and less strongly altered miRNAs. Notably, RNU6B, which is the most frequently used reference gene for miRNA studies, gave inaccurate normalization. The combination of four (miR-101, miR-125a-5p, miR-148b, and miR-151-5p) or three (miR-148b, miR-181b, and miR-874,) reference miRNAs is recommended for normalization.

Citation: Ratert N, Meyer H-A, Jung M, Mollenkopf H-J, Wagner I, et al. (2012) Reference miRNAs for miRNAome Analysis of Urothelial Carcinomas. PLoS ONE 7(6): e39309. doi:10.1371/journal.pone.0039309

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Competing Interests: The authors have declared that no competing interests exist.

* E-mail: klaus.jung@charite.de

Introduction

MicroRNAs (miRNAs) belong to a class of small noncoding RNAs of 19 to 24 nucleotides that are known to regulate signaling pathways for various cell functions. Not surprisingly, changes in miRNA expression have been associated with several diseases, including cancer [1,2]. It has been shown that different tumors have specific miRNA expression profiles and that miRNA profiles correlate with patient diagnosis, prognosis, and responses to treatment [3]. Thus, analyzing the differential expression of the microRNAome [4], defined as the entirety of all miRNAs in a cell, is of scientific and practical significance.

Several methods such as bead-based flow cytometry, microarray, deep sequencing, and real-time quantitative PCR (RT-qPCR) allow fast, high-throughput, and sensitive profiling of miRNAs. RT-qPCR

produces specific, sensitive, and reproducible quantification of nucleic acids. To overcome experimental variations in RT-qPCR analyses (RNA isolation, cDNA synthesis, PCR runs), relative quantification of miRNAs of interest based on the normalization to reference genes is the approach of choice to prevent errors within a dataset [5]. This approach complies with normalization procedures used in mRNA expression studies and is summarized in the recent MIQE guidelines [6]. Suitable reference genes should be expressed constitutively and be independent of biological changes, diseases or treatments. The use of multiple rather than single reference genes has been recommended for RT-qPCR data normalization [7,8]. The computational programs geNorm [9] and NormFinder [10] are based on this principle. These tools allow identifying the most stable reference genes from a panel of putative reference genes for normalization. Moreover, several studies in addition to our own

experiments have shown that the use of inappropriate reference genes in the relative quantification of gene expression can result in biased expression profiles [11–13]. As there are no universal reference genes [14,15], it is strongly recommended that researchers test for the most suitable reference genes specific to the tissues and experimental conditions used.

Because of our general interest in miRNAs in urological tumors and the increasing incidence of urothelial cancer [16], we performed a literature search in PubMed. The MeSH term “microRNAs” was combined with the search string [“microRNAs” OR “microRNA” OR “miRNA” OR “miRNAs”] and in combination with the MeSH term “urinary bladder neoplasms”. Fifty-eight articles published until May 2012 were identified, of which 27 investigated miRNA expression. Specifically, 20 publications reported miRNA expression by RT-qPCR and used small nuclear, nucleolar or ribosomal RNAs as well as mRNAs for normalization, namely RNU6B (15 times) [17–31], RNU48 (6 times) [17,18,32–35], RNU43 (1 time) [30], RNU44 (1 time) [18], beta-actin (1 time) [36], and 18srRNA (1 time) [32] without confirming their validity for normalization. Thus, no systematic study has been performed to identify suitable miRNA reference genes in urothelial carcinoma while corresponding studies have been performed for other cancer entities, including urological tumors [13,37–40].

Bladder cancer is the fourth most common cancer in Western industrialized countries [16]. Approximately 90% of all urothelial neoplasms are classified as urothelial carcinoma. Although surgical techniques and treatments have improved over time, bladder cancer is still a common cancer with a high mortality. To date, mechanisms of urothelial carcinogenesis have not been fully elucidated. However, miRNA expression patterns have been linked to clinical outcomes in urothelial carcinoma [18,24]. Therefore, single miRNA biomarkers or biomarker signatures of multiple miRNAs may improve risk stratification of patients and may supplement the histological diagnosis of urological tumors including bladder cancer [24,41–43]. In addition, miRNAs and their regulated genes represent interesting drug targets because miRNAs can influence the expression of multiple genes and thereby affect numerous points in disease pathways [22,44–46]. The significance of miRNAs in the regulation of signal transduction in bladder cancer was recently summarized [47]. Improved knowledge in this field will contribute to enhanced prognosis and selection of treatment strategies. However, as mentioned above, accurate quantification of miRNA expression by RT-qPCR and thus reliable expression data require proper normalization strategies. Computer programs based on various algorithms including geNorm [9], NormFinder [10], and BestKeeper [48] have been developed to rank putative reference genes according to their expression stability and indicate the best reference gene or combination of reference genes for accurate normalization.

In the present study, we aimed to systematically identify suitable reference genes for normalizing RT-qPCR assays of miRNA expression in urothelial carcinoma tissue. Using miRNA microarray analyses, we first identified invariant miRNAs that showed stable expression in both nonmalignant and malignant bladder tissue samples as candidate reference miRNAs. RT-qPCR analyses were subsequently performed for validating these miRNAs from the microarray experiments and the above mentioned small RNAs RNU6B, RNU48, and Z30 as putative reference genes. Appropriate reference miRNAs were identified by geNorm, NormFinder, and BestKeeper, and the results of unsuitable normalization are illustrated with invalid normalizers.

Materials and Methods

Patients and Tissue Samples

All bladder cancer patients went through radical cystectomy or transurethral resection at the University Hospital Charité in Berlin between 2008 and 2009 and gave written informed consent for the use of representative tissue specimens for research purposes. The study was approved by the Ethic Committee of the University Hospital Charité (File: EA1/153/07). The samples were collected immediately after surgery in liquid nitrogen and stored at -80°C until further analysis. Tumor staging was performed in conformity with the International Union Against Cancer [49] and histological grading in accordance with the WHO/ISUP criteria of 2004 [50]. In total, 58 urothelial samples were included in this study. Seventeen samples were from nonmalignant bladder tissue (15 male, 2 female patients; median age 68, range 47–80 years), 20 samples were from low-grade papillary urothelial carcinoma (18 male, 2 female patients; median age 68, range 50–86 years), and 21 samples were from high-grade tumors (14 male, 7 female patients; median age 73, range 48–82 years).

Isolation of RNA and Characterization of Quantity and Quality

Frozen histologic sections were prepared, stained with hematoxylin/eosin, and examined by uropathologists (A.E., E.K.). Only tissue specimens with more than 80% tumor cells were included in the study as tumor samples. Tissue cryotome sections (approximately 20–30 mg of tissue, wet weight) were treated with 350 μl of lysis buffer and total RNA including miRNAs was isolated using the miRNeasy Mini Kit (Qiagen, Hilden, Germany) with 30 to 50 μl of elution buffer according to the manufacturer’s protocol. An additional DNase I digestion step on the RNA binding silica gel membrane of the spin column was performed. RNA concentration and the 260 nm to 280 nm absorbance ratios were measured on the NanoDrop 1000 spectrophotometer (NanoDrop Technologies, Wilmington, DE, USA). The quality of isolated RNA was determined by the RNA integrity number (RIN) with a Bioanalyzer 2100 (Agilent Technologies, Santa Clara, CA, USA). Only samples with RIN values >5 were used. The RNA samples (medians: 693 ng/ μl ; 830 ng/mg tissue) isolated from nonmalignant as well as from low-grade and high-grade tumor tissue samples showed comparable median 260/280 absorbance ratios (2.02, 2.03, and 2.03) and RIN values (6.7, 5.9, and 6.3; Kruskal-Wallis test, $P=0.171$).

Microarray-based miRNA Analysis

Microarray analyses of eight samples each from nonmalignant tissue and low and high grade tumor specimens were performed. One-color hybridizations on human catalog 8-plex 15 K microRNA microarrays (AMADID 019118) from Agilent encoding probes for 723 human and 76 viral microRNAs from the Sanger database v10.1 were used. All reaction steps were carried out as previously described in detail [51]. After hybridization, microarrays were washed, scanned, and processed according to the supplier’s protocol. The raw data were normalized using GeneSpring GX11 Software (Agilent) with default parameters (threshold raw signal to 1.0, percent shift to 90th percentile as normalization algorithm and no baseline transformation). All microarray data have been deposited in the NCBI GEO database with accession number GSE36121. Further data analysis is described in the Results section.

Quantitative Real-time PCR

RT-qPCR analyses of miRNAs were carried out with TaqMan microRNA assays (Applied Biosystems, Foster City, CA, USA) according to the manufacturer's protocol and the MIQE guidelines [6] (Table S1) as previously described [13]. The reverse transcription of miRNAs from total RNA (10 ng) was performed with miRNA-specific stem-loop primers, 10 nmol dNTP mix, 2.6 U of RNase Inhibitor, 33.5 U of MultiScribe RT enzyme, and 1 × RT Buffer (Applied Biosystems). The generated cDNAs were stored at 20°C until analysis. The qPCR measurements were executed in white 96-well PCR plates (cat.no. 04729692001 with sealing foils) with a 10 µl of final volume containing 1 µl of RNA-specific cDNA, 1× TaqMan Universal PCR Master Mix No AmpErase UNG, and gene-specific TaqMan miRNA real-time PCR assay solution on the Light Cycler 480 Instrument (Roche Diagnostics GmbH, Mannheim, Germany; software version 1.5.0) (Table S2). The reaction was performed at 95°C for 10 min, followed by 45 cycles of 95°C for 15 s, and 60°C for 60 s. All samples were measured in triplicate; each PCR run included a no-template control and two inter-plate calibrators. All no-template controls were negative. To assess the specific amplification efficiencies, we created calibration curves from dilution series of miRNA-specific cDNAs or small nuclear RNAs (Methods S1). The efficiency was determined from the slope of the log regression plot of Cq values versus log input of cDNA. Efficiencies were in the range between 81% and 88%. All data were corrected to the PCR efficiency and to the inter-run calibrators. For that purpose, the software qBase^{PLUS} (Biogazelle NV, Zwijnaarde, Belgium) was used, employing a generalized and universally applicable quantification model based on efficiency correction, error propagation, and multiple reference gene normalization [52]. The intra-run precision for the finally considered candidate reference miRNAs miR-29c, miR-101, miR-125a-5p, miR-148b, miR-151-3p, miR-151-5p, miR-181a, miR-181b, miR-324-3p, miR-424, and miR-874 as well as the investigated small nuclear RNU6B, RNU48, and Z30 ranged from 0.15% to 0.35% for mean Cq values between 21.93 to 26.65. The between-run precision (n = 42) measured for the control miR-21 was found to be 1.62% (mean Cq ± standard deviation: 28.35 ± 0.46).

Data Analysis

Statistical analyses were performed using GraphPad Prism Version 5.04 (GraphPad Software Inc., La Jolla, CA, USA). Non-parametric tests (Mann-Whitney U test; Kruskal-Wallis test with Dunn's multiple comparison test) were used to analyze significant differences between independent groups. The Spearman correlation coefficients were applied to calculate the relationships between the miRNAs as well as between the clinical variables and the expression of candidate reference miRNAs. P values <0.05 (two-tailed) were considered statistically significant.

The assessment of the putative reference genes for normalization was evaluated by the computer programs geNorm [9] using the improved version geNorm^{Plus} as an implementation of the software qBase^{PLUS} (Biogazelle, Belgium) [52], NormFinder [10], and BestKeeper [48].

Results

Selection of Candidate Reference miRNAs by Microarray Analysis

To identify putative reference miRNAs in the miRNA microarray data obtained from the eight samples of each tissue group, the following criteria were used: (a) miRNAs had to be detected in Genespring GX11 software as "present" in all

Table 1. Candidate reference miRNAs selected from microarray analysis.[†]

miRNA ^{&}	ID according miRBase version [#]	Selection criterion [§]
<i>hsa-miR-15a</i>	hsa-miR-15a (v10.1)	R
	hsa-miR-15a-5p (v18)	
<i>hsa-miR-20b</i>	hsa-miR-20b (v10.1)	R
	hsa-miR-20b-5p (v18)	
<i>hsa-miR-29c</i>	hsa-miR-29c (v10.1)	R
	hsa-miR-29c-3p(v18)	
<i>hsa-miR-101</i>	hsa-miR-101 (v10.1)	N
	hsa-miR-101-3p (v18)	
<i>hsa-miR-107</i>	hsa-miR-107 (v10.1)	N
	hsa-miR-107 (v18)	
<i>hsa-miR-125a-5p</i>	hsa-miR-125a-5p (v10.1)	N
	hsa-miR-125a-5p (v18)	
<i>hsa-miR-148b</i>	hsa-miR-148b (v10.1)	R
	hsa-miR-148b-3p (v18)	
<i>hsa-miR-151-3p</i>	hsa-miR-151-3p (v10.1)	R
	hsa-miR-151-3p (v18)	
<i>hsa-miR-151-5p</i>	hsa-miR-151-5p (v10.1)	N
	hsa-miR-151-5p (v18)	
<i>hsa-miR-181a</i>	hsa-miR-181a (v10.1)	R
	hsa-miR-181a-5p (v18)	
<i>hsa-miR-181b</i>	hsa-miR-181b (v10.1)	R
	hsa-miR-181b-5p (v18)	
<i>hsa-miR-324-3p</i>	hsa-miR-324-3p (v10.1)	N
	hsa-miR-324-3p (v18)	
<i>hsa-miR-424</i>	hsa-miR-424 (v10.1)	N
	hsa-miR-424-5p (v18)	
<i>hsa-miR-513a-5p</i>	hsa-miR-513a-5p (v10.1)	R
	hsa-miR-513a-5p (v18)	
<i>hsa-miR-874</i>	hsa-miR-874 (v10.1)	N
	hsa-miR-874 (v18)	
<i>hsa-miR-939</i>	hsa-miR-939 (v10.1)	N
	hsa-miR-939 (v18)	

[†]The TaqMan MicroRNA Assay ID, miRBase accession number, and the sequence for each miRNA are compiled in Table S2.

[&]miRNAs marked in Italics were not included in further analyses because their low expression level was beyond the dynamic range of the assay (>35Cq) (further details see text).

[#]The miRNA ID from the miRBase version 10.1 and 18, respectively.

[§]Symbols "N" and "R" indicate the selection of the candidate reference miRNAs based on normalized or raw microarray data as described in the text. doi:10.1371/journal.pone.0039309.t001

examined 24 samples to filter out signals that did not reach a minimum of intensity, (b) the absolute fold change between the nonmalignant and the two cancerous groups had to be lower than 1.2-times with (c) no significant differences (P>0.05) between the groups. Based on the total of 723 human miRNA species located on the Agilent microarray chip according to the miRBase version 10.1, we identified 101 miRNAs that were flagged as "present" in all of the examined groups (Table S3). Eight of these miRNAs showed absolute fold changes lower than 1.2-times and had no significant differences between the groups (Table 1, indicated by

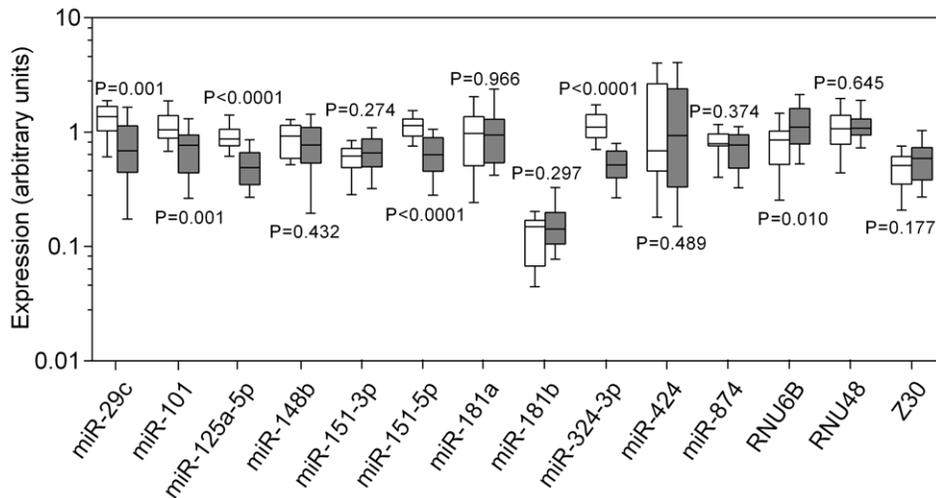


Figure 1. Expression of candidate reference genes in human nonmalignant and malignant bladder tissue samples. RT-qPCR analyses were performed from 17 nonmalignant bladder tissue samples and 41 samples from low-grade and high-grade papillary urothelial carcinoma. Expression levels of the candidate reference genes are given as arbitrary units. Boxes (blank, nonmalignant samples; black, malignant samples) represent lower and upper quartiles with median as horizontal line; whiskers depict the 10 and 90 percentiles. Significances are illustrated as *P* values of the Mann-Whitney *U* test.
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the symbol “N”). To avoid normalization artifacts of the microarray data, we also used raw microarray expression data. Thus, with the criteria mentioned above, we revealed a second set of eight candidate reference miRNAs (Table 1, indicated by symbol “R”). Taking these sets together, 16 putative reference miRNAs were included in further analyses (Table 1; Table S2).

Validation of Candidate Reference Genes by RT-qPCR

To increase the statistical power to find suitable reference miRNAs, in addition to the 24 analyzed samples in the microarray experiments, we included nine nonmalignant and 25 malignant tissue samples as described in the section “Patients and tissue samples” to validate the aforementioned 16 candidate reference miRNAs in more detail by RT-qPCR. Furthermore, the set of candidate reference miRNAs was extended by the small RNAs RNU6B, RNU48, and Z30 that were commonly used for expression normalization in the literature as stated in the Introduction. First, to determine if reliable quantification of these putative normalizers is feasible by RT-qPCR, three RNA pools were prepared containing equal amounts of RNA from the samples used in the microarray analysis. miR-15a, miR-20b, miR-107, miR-513a-5p, and miR-939 showed *C_q* values >35 in the pools and were excluded from further analyses because accurate quantification would be questionable. By this preselection, 11 putative reference miRNAs (Table 1: miR-29c, miR-101, miR-125a-5p, miR-148b, miR-151-3p, miR-151-5p, miR-181a, miR-181b, miR-324-3p, miR-424, and miR-874) as well as RNU6B, RNU48, and Z30 were further investigated and showed *C_q* values ranging from 22 (RNU48) to 28 (miR-324-3p).

In the second step, all 14 reference candidates were separately measured in the 58 samples (Figure 1). The expression levels significantly differed for miR-29c ($P=0.0012$), miR-101 ($P=0.0007$), miR-125a-5p ($P<0.0001$), miR-151-5p ($P<0.0001$), miR-324-3p ($P<0.0001$), and RNU6B ($P=0.0101$) between nonmalignant and malignant samples. The remaining eight miRNAs, namely miR-148b, miR-151-3p, miR-181a, miR-181b, miR-424, miR-874, RNU48, and Z30, revealed no significant differences between nonmalignant and malignant samples

($P>0.05$). We followed the general recommendation of the geNorm program and included all these putative reference miRNAs and small RNAs in further analyses for reassessing their potential contribution as normalizers. However, miR-151-3p was excluded due to the fact that miR-151-3p and miR-151-5p are mature miRNAs of the same pre-miRNA and miR-151-5p exhibited higher expression in examined samples.

Association between the Candidate Reference miRNAs and Clinical Variables

The correlation between the putative reference miRNAs and the correlation of these miRNAs with age, sex, and tumor characteristics were determined. Spearman rank correlations are summarized in Table S4. Classifying miRNA pairs with coefficients ≥ 0.60 as co-expressed, we identified this characteristic co-expression feature among the four miRNAs miR-101, miR-125a-5p, miR-151-5p and miR-324-3p as well as between miR-148 and miR-151-3p, and between miR-181a and miR-181b. The correlation between miR-101, miR-151-5p, and miR-324-3p was remarkable.

The expression of the 11 miRNAs and three small RNAs was not associated with age (Spearman rank correlation from $r_s -0.23$ to 0.177, *P* values from 0.083 to 0.974), sex (Mann-Whitney *U* test, *P* values from 0.062 to 0.851), or tumor stage (Ta, T1, T2, T3; Kruskal-Wallis test, *P* values from 0.092 to 0.826, except for miR-29c, which had $P=0.044$). Differences in expression between low-grade and high-grade tumors were found for miR-29c (down-regulated, $P=0.005$), miR-874 (down-regulated, $P=0.019$), miR-181a (up-regulated, $P=0.031$), and miR-181b (up-regulated, $P=0.0008$), while all other miRNAs were not differentially expressed (*P* values from 0.092 to 0.826).

Identification of Suitable Reference miRNAs using geNorm, NormFinder, and BestKeeper

To identify suitable reference genes for the normalization of miRNA expression, we applied the aforementioned three computer programs geNorm, NormFinder, and BestKeeper. GeNorm, an implementation of the new software qBase^{PLUS}, automatically returns the average expression stability value *M* and the average

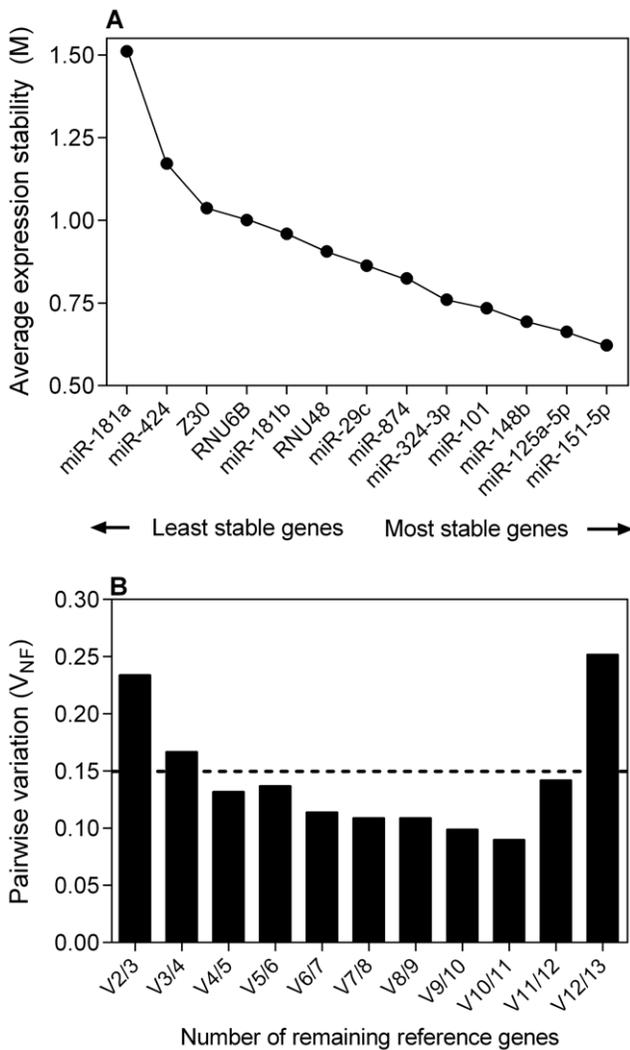


Figure 2. geNorm analysis of RT-qPCR-based candidate reference genes. (A) The geNorm analysis shows the calculation of the average expression stability value M of all candidate reference genes determined by RT-qPCR. Genes with the highest M value have the least stable expression, while the genes with the lowest M value have the most stable expression. The x-axis presents the ranking of reference genes in order of increasing stability from left to the right. (B) Calculation of the optimal number of reference genes for normalization. geNorm calculates a normalization factor assessing the optimal number of reference genes for generating that factor. The normalization factor is calculated from at least two genes taking into account the variable V as the average pairwise variation (V_{NF}) between two sequential normalization factors. The thin broken line illustrates the cut-off value $V_{NF} < 0.15$. In this experiment, the optimal number of reference genes was four (V4/5). geNorm shows the variation of the normalization factor of four genes as the best combination (miR-101, miR-148b; miR-125a-5p, and miR-151-5p) in relation to five genes as shown in (A) and in the following order. All the results are presented according to the output files of the geNorm program.
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pairwise variation V of a particular gene with all other control genes. The highest M value indicates the gene with the least stable expression. Figure 2A and Table 2 indicate the M values and the resulting ranking order of all investigated candidate reference miRNAs and small RNAs based on expression stability. miR-181a ($M = 1.511$) showed the highest M value, whereas miR-151-5p ($M = 0.622$) and miR-125a-5p ($M = 0.663$) showed the lowest M

values. Consequently, miR-181a had the least stable expression while miR-125a-5p and miR-151-5p had the most stable expression. Additionally, geNorm calculates a normalization factor (V_{NF} value), which is a criterion for the optimum number of reference genes (Figure 2B). The program recommends V_{NF} values less than 0.15 for proper normalization. When this cut-off value is achieved, it is not necessary to include any additional reference genes. As illustrated in Figure 2B, the four miRNAs miR-101, miR-148b, miR-125a-5p, and miR-151-5p (V_{NF} value 0.14) were recommended as an optimum reference miRNA set for normalization; the best combination of two reference miRNAs was miR-125a-5p and miR-151-5p, and the best single reference miRNA was miR-151-5p. After excluding the potentially deregulated reference miRNAs mentioned above, geNorm analysis was repeated. However, under these conditions, geNorm calculated a V_{NF} value > 0.15 and therefore did not recommend a normalization set.

Similar to geNorm, NormFinder identified genes with the lowest M values as the most stably expressed reference targets (Table 2). NormFinder ranked the four best reference genes for normalization as miR-148, miR-874, miR-181b, and Z30. Z30 and miR-125a-5p were recommended as the best combination, and miR-148b was recommended as the best single normalizer (Table 2).

BestKeeper considers all genes in all observed groups. First, BestKeeper determines the geometric mean and coefficient of variance. Genes with a standard deviation greater than 1 were assumed to be inconsistent. In the second step, the inter-gene relationships were examined by pairwise correlation analysis. This calculation is used to determine whether the gene expression exhibits a similar behavior. Candidate reference genes that highly correlate with each other are included in the BestKeeper-Index calculation. The program provides only an analysis of ten candidate reference genes simultaneously. Therefore, we excluded the reference targets with the lowest M values as determined by geNorm (miR-181a) and NormFinder (miR-424) and also excluded miR-29c (rank 9 by geNorm; rank 11 by NormFinder). Under these conditions, BestKeeper ranked RNU48 as the best reference gene, followed by miR-874, miR-151-5p, and Z30.

The comparison of the summarized data in Table 2 shows that the results provided by NormFinder and BestKeeper displayed slight differences from the geNorm analysis but did have some overlap. While geNorm recommended miR-101, miR-125a-5p, miR-148b, and miR-151-5p for proper normalization, NormFinder indicated miR-148b as the best reference miRNA and miR-125a-5p as a part of the best combination. Additionally, BestKeeper identified miR-151-5p within the four most stably expressed miRNAs. As stated in the Introduction, the small nucleolar RNU6B is commonly used for miRNA expression normalization and in our study was ranked 10th by geNorm, 12th by NormFinder, and 9th by BestKeeper (Table 2). Thus, RNU6B seems to be a rather inappropriate reference gene for the miRNA expression normalization in studies on bladder cancer.

Influence of Reference miRNA Selection on the Accuracy of Relative Quantification

To illustrate the impact of reference gene selection on miRNA expression analysis, we applied different normalization strategies for the relative quantification of miR-200a and miR-20a (Figure 3A–B). A preliminary evaluation of the miRNA microarray data showed a strong up-regulation of miR-200a (fold change 22.1) and a less robust, but significant up-regulation of miR-20a (fold change 2.78) in the tumor samples compared to the nonmalignant tissue samples. Different normalization approaches

Table 2. Ranking of candidate reference miRNAs and small RNAs in human nonmalignant and malignant bladder tissues according to their stability value using geNorm, NormFinder, and BestKeeper algorithms.

Gene name	geNorm		NormFinder		BestKeeper	
	Stability value ^{&}	Rank	Stability value ^{&}	Rank	SD [$\pm x$ -fold] [#]	Rank
miR-101	0.734	4	0.215	8	0.69	10
miR-125a-5p	0.663	2	0.192	6	0.62	5
miR-148b	0.693	3	0.086	1	0.65	8
miR-151-5p	0.622	1	0.230	9	0.60	3
miR-181a	1.511	13	0.209	7	–	
miR-181b	0.959	9	0.155	3	0.62	6
miR-29c	0.863	7	0.246	10	–	
miR-324-3p	0.76	5	0.291	11	0.64	7
miR-424	1.172	12	0.371	13	–	
miR-874	0.824	6	0.102	2	0.53	2
RNU6B	1.001	10	0.349	12	0.67	9
RNU48	0.906	8	0.173	5	0.51	1
Z30	1.037	11	0.171	4	0.61	4
Best gene	miR-151-5p		miR-148b		RNU48	
Best combination	miR-101, miR-125a-5p, miR-148b, miR-151-5p		Z30, miR-125a-5p		–	

[&]High expression stability is indicated by low stability value.

[#]SD [$\pm x$ -fold]: standard deviation of the absolute regulation coefficients. SD >1 can be considered inconsistent.

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were used based on the recommendations by geNorm, NormFinder, and BestKeeper as described above. As shown in Figure 3, we normalized the expression of miR-200a and miR-20a using the geNorm recommended reference miRNAs as follows: (a) the combination of four reference miRNAs that were suggested to be necessary for reliable normalization (geometric mean of miR-101, miR-125a-5p, miR-148b, and miR-151-5p; Table 2); (b) the three best ranked miRNAs according to their M values (miR-125a-5p, miR-148b, and miR-151-5p), and (c) the best combination of two miRNAs (miR-125a-5p and miR-151-5p). The NormFinder recommended approaches were the following: (d) the best two reference gene combination (miR-125a-5p and Z30); (e) the three best ranked reference miRNAs (miR-148b, miR-181b, and miR-874), and (f) the best single miRNA miR-148b. Based on the BestKeeper recommendation, we also used (g) the calculated best single reference gene RNU48. In addition, we performed normalization with (h) RNU6B, which was estimated by all three programs to have low usefulness as a reference gene but is frequently used in expression studies. Regardless of the normalization approach, miR-200a was found to be up-regulated (Figure 3A). However, the expression pattern of miR-20a was different depending on the normalization approach (Figure 3B). Using the reference miRNAs recommended by geNorm or NormFinder, miR-20a appeared to be up-regulated in tumor samples, whereas normalization with RNU6B or RNU48 as recommended by BestKeeper did not show up-regulation of this miRNA. Thus, although all reference miRNA suggestions by geNorm and NormFinder were obviously suited to be appropriate for normalization, we recommend the use of more than two reference miRNAs preferring the use of four miRNAs (miR-101, miR-148b, miR-125a-5p, and miR-151-5p) as recommended by geNorm or the combination of three miRNAs (miR-148b, miR-181b, and miR-874) suggested by NormFinder. The two-miRNA combinations or single miRNAs should be cautiously considered

as alternative normalization approaches only if limited sample material is available for analysis.

Discussion

The selection of suitable reference genes as normalizers for relative quantification of mRNA and miRNA expression is essential to avoid erroneous expression results and to improve the comparability of gene expression data between different studies. Different models such as the global mean normalization [5], panels of miRNAs [37] or small RNAs [53] have been suggested for the normalization of miRNA expression data. D'haene et al. [54] recently reported that the side-by-side comparison of small nuclear RNA normalization with global mean normalization indicated that small nuclear RNAs are less efficient in reducing the technical variation and do not reveal in accurate expression differences. In addition, the recommended global mean normalization method [5] that is also included in the algorithm of the qBase^{PLUS} software requires a large number of genes, for example in microarray, deep sequencing, bead-based or TaqMan array card analyses. Thus, the global normalization approach is not feasible in RT-qPCR studies because only a few miRNAs are generally measured. In this case, the normalization of miRNA RT-qPCR data with suitable miRNA reference markers can be considered as the method of choice [55].

Studies to identify and validate suitable reference miRNAs have been performed for several cancers [13,37–40]. As discussed in the Introduction, it is therefore quite astonishing that no miRNA expression studies in bladder cancer have used endogenous miRNAs for normalization. Only nuclear, nucleolar, and ribosomal RNAs as well as mRNAs have been used. However, the different lengths of these RNAs compared to miRNAs result in different physico-chemical properties with different isolation efficiencies and degradation [56,57]. miRNAs are more stable

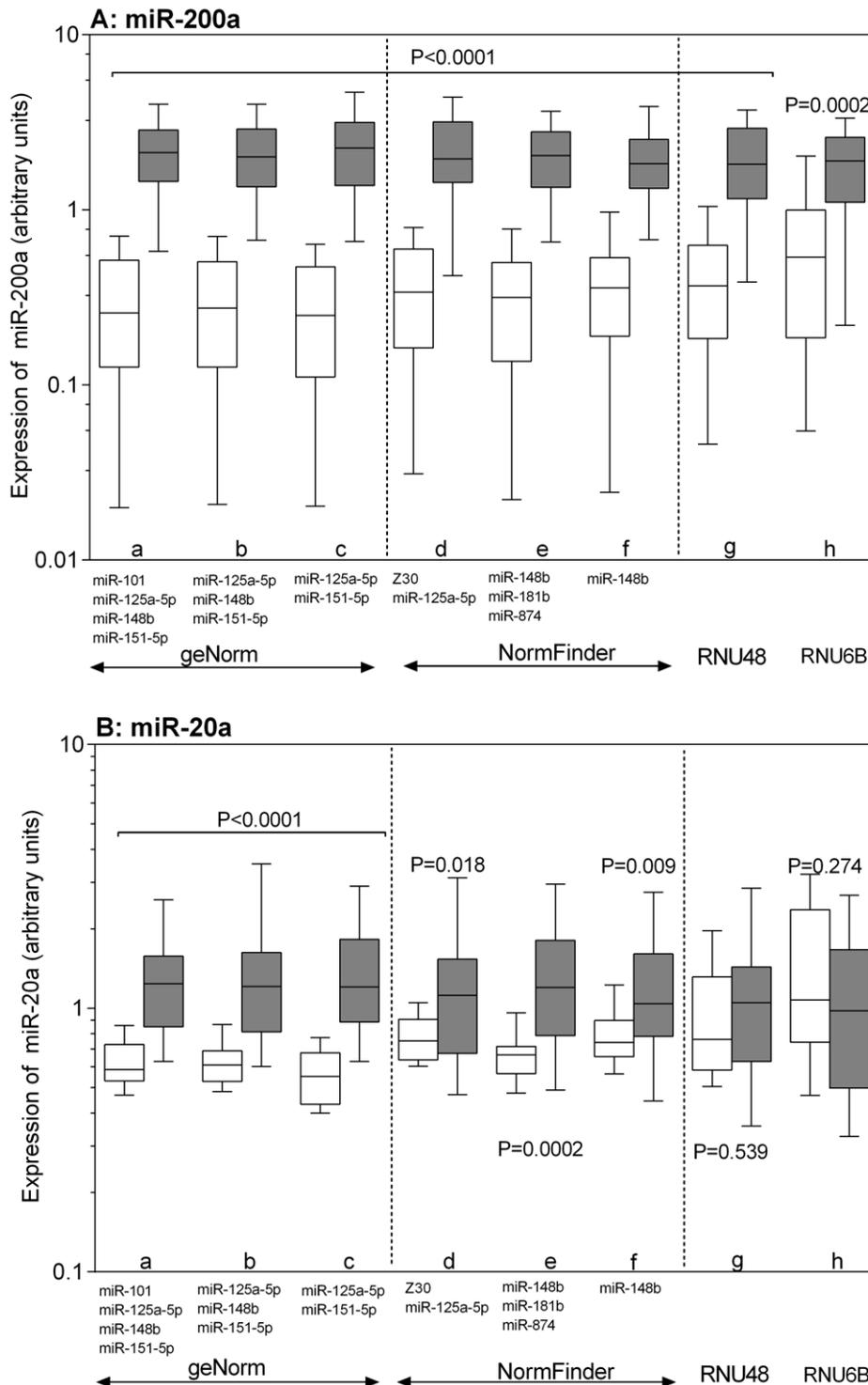


Figure 3. Effects of different normalization approaches on the expression of miR-200a and miR-20a. The relative expression of (A) miR-200a and (B) miR-20a as highly and moderately differentially expressed miRNAs, respectively was calculated using the following normalization strategies recommended by geNorm (a–c), NormFinder (d–f), BestKeeper (g), and RNU6B (h). The geNorm approaches were: (a) the four-reference-miRNA combination recommended as necessary number of reference miRNAs (miR-101, miR-125a-5p, miR-148b, miR-151-5p); (b) the three best ranked miRNAs according to their M values (miR-125a-5p, miR-148b, and miR-151-5p) and (c) the best two-gene-reference combination (miR-125a-5p, miR-151-5p). NormFinder normalization approaches were: (d) the best two reference gene combination (miR-125a-5p, Z30); (e) the three best ranked reference genes (miR-148b, miR-181b, miR-874); (f) the best single miRNA, miR-148b. BestKeeper normalization approach was (g) RNU48; (j) RNU6B as the most frequently recommended normalizer in bladder cancer studies. Values are given as arbitrary units; boxes (blank, nonmalignant tissue; black, malignant tissue) represent lower and upper quartiles with medians as horizontal line; whiskers depict the 10–90 percentiles. Significances are illustrated as P values of the Mann-Whitney U test.
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than mRNAs or small RNAs like RNU6B, and they can therefore be more accurately detected in tissues [57]. In addition, different techniques of reverse transcription used for miRNAs and the other RNAs make the latter less suitable for normalization. Furthermore, as previously shown for mRNAs, the tissue-specific expression of miRNAs is also reflected in the behavior of putative endogenous reference genes [14,15]. Thus, RT-qPCR-based miRNA expression studies optimally require normalization by reference miRNAs. Previous reports from our group have demonstrated the importance of suitable reference miRNAs in avoiding biased results in miRNA expression studies in other urological tumors [13,39]. These data strongly support the need for determining appropriate endogenous reference miRNAs to allow stringent normalization of miRNA expression patterns in urothelial carcinoma.

To the best of our knowledge, the present study is the first systematic investigation of suitable normalizers for relative quantification of miRNA expression in bladder cancer. In this study, we combined different strategies for identifying suitable reference miRNAs. A four step approach was used. First, to obtain an overview of the miRNAome in bladder cancer tissue, miRNA microarray analyses for nonmalignant and malignant bladder samples were performed to identify invariant miRNAs as stably-expressed candidate reference miRNAs within the data set. Second, these candidate reference miRNAs were validated by RT-qPCR, in addition to RNU6B, RNU48, and Z30, the most frequently reported reference genes for miRNA expression studies in bladder cancer. Third, the statistical algorithms geNorm, NormFinder, and BestKeeper were applied to identify the most useful endogenous reference miRNAs for relative quantification. Finally, the impact of different normalization approaches was illustrated for two deregulated miRNAs in bladder cancer tissue to emphasize the importance of an appropriate normalization approach.

The starting point of the present study was miRNA microarray analysis. According to the criteria for the microarray data evaluation and the measurability criterion for subsequent RT-qPCR analysis (C_q values <35), 11 invariant miRNAs were identified to be putative reference miRNAs. Since miR-151-3p and miR-151-5p derive from the same pre-miRNA and miR-151-5p exhibited slightly higher expression in examined samples, we included only miR-151-5p in further analysis. A data search in the miRNA bladder cancer studies mentioned in the Introduction showed that miR-29c, one of these 11 invariant, stably expressed miRNAs, was found to be down-regulated in two microarray studies [19,36]. Our subsequent RT-qPCR confirmed this observation (Figure 1). Although we did not eliminate this miRNA from the subsequent analysis for the validation of suitable reference miRNAs through geNorm, NormFinder, and BestKeeper, miR-29c was never recommended as a reference miRNA by one of these evaluation tools in our following analyses. This finding also indicates the usefulness of the software packages in the search for suitable reference genes.

The putative reference miRNAs identified by the microarray analyses, except miR-151-3p as mentioned, were included with the additional RNAs RNU6B, RNU48, and Z30 in the geNorm, NormFinder, and BestKeeper analysis. Differences in expression observed in the subsequent RT-qPCR measurements between nonmalignant, low-grade, and high-grade tumor samples as well as co-expressions of genes did not exclude candidate reference genes. However, as comprehensively described in the Results section, geNorm, NormFinder, and BestKeeper did not always recommend the same reference miRNAs for normalization (Table 2). The lack of agreement between geNorm and NormFinder results has been described previously [15]. The reasons for these

differences in the ranking order of the putative reference miRNAs might be due to the different calculation models on which the tools are based. NormFinder is an ANOVA-based model, geNorm uses a pairwise comparison model, and BestKeeper determines the optimal reference genes by employing the pairwise correlation analysis on all pairs of candidate reference genes. While the geNorm approach is theoretically robust with regard to inter-sample variations arising from sources such as differing RNA input and quality, it has been shown to prefer co-regulated genes in the selection as normalizers [10]. In this study, geNorm also recommended co-regulated reference miRNAs (miR-101 with miR-125a-5p, miR-151-5p) but miR-324-3p was never recommended as normalizer despite its strong correlation with miR-101 and miR-151-5p.

The differences between the recommended reference miRNAs by the three programs prompted us to validate their suitability in clinical samples (Figure 3A–B). The importance of selecting suitable reference genes for accurate miRNA expression data has been shown not only in mRNA but also in miRNA expression studies [13,37,39,40]. We tested the suitability of the different approaches with miR-200a, a highly up-regulated miRNA, and miR-20a, which is up-regulated less robustly (Figure 3A, B). The results clearly demonstrated that RNU6B, which is the most frequent normalizer used in previous miRNA expression studies in bladder cancer, and RNU48, which was recommended by BestKeeper, were unable to confirm the small expression changes, e.g. for miR-20a. The poor quality of RNU6B as a reference gene has already been reported in miRNA expression studies in renal cell carcinoma and prostate cancer [13,58], where its altered expression stability depended on the degradation of the RNA as compared with miRNAs [13]. In contrast, all geNorm and NormFinder recommendations for single and multiple reference miRNA combinations proved to be suitable normalization approaches in the present study for revealing not only strongly but also less robustly deregulated miRNA expression levels between nonmalignant and malignant urothelial tumor samples. However, we recommend the combination of four (miR-101, miR-125a-5p, miR-148b, and miR-151-5p) or three (miR-148b, miR-181b, and miR-874) reference miRNAs. Although the normalization with the best single (NormFinder) or the best two (geNorm) reference miRNAs in our study gave comparable results to the larger gene sets, the use of multiple reference miRNAs is critical in achieving more reliable expression data [7–10].

In summary, the present study was the first systematic investigation to identify suitable reference miRNAs in a transparent and comprehensive manner for the relative quantification of the microRNAome in urothelial carcinoma. It was based on a four-step approach with microarray analyses, RT-qPCR validation, reference miRNA selection through computer software, and proof of principle with different miRNA expression levels. Starting with 16 putative reference miRNAs from the microarray analysis and three additional small RNAs from the literature, we validated several combinations of reference miRNAs for miRNA expression studies in bladder cancer. We believe that these are robust methods that will allow future studies on the functional roles of miRNAs as regulators in signal transduction and metabolic pathways that are associated with small expression changes.

Supporting Information

Table S1 Description of the experimental details of the RT-qPCR analyses according to the checklist of the MIQE guidelines.

(PDF)

Table S2 TaqMan assays for microRNAs and small nuclear and nucleolar RNAs.

(PDF)

Table S3 List of the 101 miRNAs from the microarray for the identification of candidate reference miRNAs.

(PDF)

Table S4 Spearman rank correlation coefficients (r_s) between the candidate reference genes.

(PDF)

Methods S1 qPCR validation experiments according to the MIQE guidelines with respect to the calibration curves and the dynamic range of measurements.

(PDF)

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Author Contributions

Conceived and designed the experiments: NR MJ HJM AE SW KJ. Performed the experiments: NR MJ HJM IW EK AE. Analyzed the data: NR MJ HAM KJ. Contributed reagents/materials/analysis tools: HJM HAM. Wrote the paper: NR HAM KJ. Provided cancer specimens and clinical information: KM EK AE SW. Critical revision and final approval of manuscript: NK HAM MJ HJM IW KM EK AE SW KJ.

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Ergänzende Informationen: Reference miRNAs for miRNAome analysis of urothelial carcinomas

Die folgenden Seiten beinhalten die ebenfalls veröffentlichte "Supporting information" zur eingereichten Publikation „Reference miRNAs for miRNAome analysis of urothelial carcinomas“.

Supporting Information

Table S1 Description of the experimental details of the RT-qPCR analyses according to the checklist of the MIQE guidelines. The items in the checklist summarize the characteristics of the RT-qPCR analyses according to Bustin et al. (Clin Chem (2009) 611-622) and indicate where the details are described in the manuscript.

Table S2 TaqMan assays for microRNAs and small nuclear and nucleolar RNAs. TaqMan assays from Applied Biosystems for the examined mature miRNAs and small nuclear RNAs are listed. The assay name, assay ID, miRBase accession number, miRBase ID, and the sequence are given for the miRNAs and correspondingly for the small nuclear and nucleolar RNAs. miRNAs were identified by the permanently assigned miRBase accession number, the miRBase-prescribed ID related to the miRBase version and the sequence.

Table S3 List of the 101 miRNAs from the microarray for the identification of candidate reference miRNAs. All miRNAs of the microarray are listed that were flagged as "present" in all examined 24 tissue samples by the Genespring GX11 software.

Table S4 Spearman rank correlation coefficients (r_s) between the candidate reference genes

Methods S1 qPCR validation experiments according to the MIQE guidelines with respect to the calibration curves and the dynamic range of measurements. Calibration curves were generated with diluted cDNAs. The Cq values were calculated automatically by the LightCycler software, release 1.5.0 using the "second derivative maximum" method. The slopes, intercepts, and errors of the regression lines of the calibration curves from these dilution series and the PCR efficiencies ($E=10^{-1/\text{slope}}$) including the dynamic range and the Cq variation at the lower limit (the endpoint of the linear dynamic range) were also calculated by this software. Data and curves of the miR-26a, miR-151-3p, and RNU48 are exemplarily shown as follows. Since efficiencies did only differ in the second decimal place confirming the manufacturer's information that the TaqMan miRNA assays run with equivalent amplification efficiencies, we used the data of the calibration curve of miR-26a for all miRNAs and those of the curve of RNU48 for all RNUs.

Table S1. Description of the experimental details of the RT-qPCR analyses according to the checklist of the MIQE guidelines. The items in the checklist summarize the characteristics of the RT-qPCR analyses according to Bustin et al. [Clin Chem (2009) 611-622] and indicate where the details are described in the manuscript.

ITEM TO CHECK	IMPOR- TANCE ^{&}	CHECK- LIST	WHERE IN THE MANUSCRIPT; ADDITIONAL COMMENT
EXPERIMENTAL DESIGN			
Definition of experimental and control groups	E	Yes	Materials and Methods: Patients and tissue samples.
Number within each group	E	Yes	Materials and methods: Patients and tissue samples.
Assay carried out by core lab or investigator's lab?	D	Yes	All assays were performed in investigator's lab.
Acknowledgement of authors' contributions	D	Yes	Contributors who do not meet the authorship as defined by the journal are listed in the Acknowledgement section.
SAMPLE			
Description	E	Yes	Materials and Methods.
Volume/mass of sample processed	D	Yes	Materials and Methods: Isolation of RNA....
Microdissection or macrodissection	E	Yes	Materials and Methods: Isolation of RNA....; macrodissections with histological verification.
Processing procedure	E	Yes	Materials and Methods: Patients and tissue samples; Isolation of RNA...
If frozen – what and how quickly?	E	Yes	Materials and Methods: Patients and tissue samples.
If fixed – with what, how quickly?	E	N/A	-
Samples storage conditions and duration (esp. for FFPE samples)	E	Yes	Materials and Methods: Patients and tissue samples.
NUCLEIC ACID EXTRACTION			
Procedure and/or instrumentation	E	Yes	Materials and Methods: Isolation of RNA and characterization of quantity and quality; references are indicated.
Name of kit and details of any modifications	E	Yes	Materials and Methods: Isolation of RNA and characterization of quantity and quality; references are indicated.
Source of additional reagents used	E	Yes	RNase-free DNase set; Qiagen (cat.no. 79254); see subsequent information.
Details of DNase or RNase treatment	E	Yes	Materials and Methods: Isolation of RNA....; an optional on column digestion DNase step was used.
Contamination assessment (DNA or RNA)	E	Yes	See previous comment; according to Chen et al. (Nucleic Acids Res 33(2005) e179) miRNA measurements by the TaqMan assays are not affected by genomic DNA; see also comment on Cqs with and without RT.
Nucleic acid quantification	E	Yes	Materials and Methods: Isolation of RNA and characterization of quantity and quality; references are indicated.
Instrument and method	E	Yes	Materials and Methods: Isolation of RNA and characterization of quantity and quality; references are indicated.
Purity (A260/A280)	D	Yes	Materials and Methods: Isolation of RNA and characterization of quantity and quality; references are indicated.
Yield	D	Yes	Materials and Methods: Isolation of RNA and characterization of quantity and quality.
RNA integrity method/instrument	E	Yes	Materials and Methods: Isolation of RNA and characterization of quantity and quality; Bioanalyzer 2100 (Agilent).
RIN/RQI or Cq of 3' and 5' transcripts	E	Yes	RIN values; given in Materials and Methods: Isolation of RNA and characterization of quantity and quality.
Electrophoresis traces	D	No	-
Inhibition testing (Cq dilutions, spike or other)	E	Yes	Dilution experiments were performed; PCR efficiencies were found 81%-88%; see also Supporting Information, Methods S1. For all clinical samples, identical isolation procedures were performed.
REVERSE TRANSCRIPTION			
Complete reaction condition	E	Yes	Materials and Methods: Quantitative real-time PCR; references are indicated.
Amount of RNA and reaction volume	E	Yes	Materials and Methods: Quantitative real-time PCR; references are indicated.

Priming oligonucleotide (if using GSP) and concentration	E	Yes	Materials and Methods: Quantitative real-time PCR; references are indicated.
Reverse transcriptase and concentration	E	Yes	Materials and Methods: Quantitative real-time PCR; references are indicated.
Temperature and time	E	Yes	Materials and Methods: Quantitative real-time PCR; references are indicated.
Manufacturer and reagents and catalogue numbers	D	Yes	Materials and Methods: Quantitative real-time PCR; references are indicated.; see also Supporting Information, Table S2.
Cqs with and without RT	D*	Yes	There were no Cqs < 40 in reactions without RT.
Storage conditions of cDNA	D	Yes	Materials and Methods: Quantitative real-time PCR; - 20°C.
qPCR TARGET INFORMATION			
If multiplex, efficiency and LOD of each assay	E	N/A	-
Sequence accession number	E	Yes	Table 1 and Supporting Information, Table S2.
Location of amplicon	D	Yes	Use of miRNAs specific TaqMan assays; specificity guaranteed by the manufacturer.
Amplicon length	E	N/A	Use of miRNAs specific TaqMan assays; specificity guaranteed by the manufacturer.
<i>In silico</i> specificity screen (BLAST, etc.)	E	N/A	Use of miRNA specific TaqMan assays; specificity guaranteed by the manufacturer.
Pseudogenes, retropseudogenes or other homologs?	D	N/A	Use of miRNA specific TaqMan assays; specificity guaranteed by the manufacturer.
Sequence alignment	D	N/A	Use of miRNA specific TaqMan assays; specificity guaranteed by the manufacturer.
Secondary structure analysis of amplicon	D	N/A	Use of miRNA specific TaqMan assays; specificity guaranteed by the manufacturer.
Location of each primer by exon or intron (if applicable)	E	Yes	Use of miRNA specific TaqMan assays; specificity guaranteed by the manufacturer.
What splice variants are targeted?	E	Yes	Specificity guaranteed by manufacturer of the TaqMan assays.
qPCR OLIGONUCLEOTIDES			
Primer sequences	E	N/A	The manufacturer does not provide this information for miRNAs; see also Supporting Information, Table S2.
RTPrimerDB Identification Number	D	N/A	Use of miRNA specific TaqMan assays; see also Supporting Information, Table S2.
Probe sequences	D**	N/A	The manufacturer does not provide this information for miRNAs; see also Supporting Information, Table S2.
Location and identity of any modifications	E	Yes	The manufacturer does not provide this information for miRNAs; see also Supporting Information, Table S2.
Manufacture of oligonucleotides	D	Yes	Applied Biosystems as part of Life Technologies.
Purification method	D	No	Applied Biosystems does not provide information.
qPCR PROTOCOL			
Complete reaction conditions	E	Yes	Materials and Methods: Quantitative real-time PCR; references are indicated; use of miRNA specific TaqMan assays.
Reaction volume and amount of cDNA/DNA	E	Yes	Materials and Methods: Quantitative real-time PCR; references are indicated.
Primer, (probe), Mg ⁺⁺ and dNTP concentration	E	Yes	Materials and Methods: Quantitative real-time PCR; references are indicated; use of miRNA specific TaqMan assays.
Polymerase identity and concentration	E	Yes	Materials and Methods: Quantitative real-time PCR; references are indicated; use of miRNA specific TaqMan assays.
Buffer/kit identity and manufacture	E	No	Materials and Methods: Quantitative real-time PCR; references are indicated; use of miRNA specific TaqMan assays.
Exact chemical constitution of the buffer	D	Yes	The manufacturer does not provide this information.
Additives (SYBR Green I, DMSO, ect.)	E	Yes	Use of TaqMan assays without additional additives.
Manufacturer of plates/tubes and catalog number	D	Yes	Materials and Methods: Quantitative real-time PCR; white 96-well PCR plates (Roche; cat.no. 04729692001 with sealing foils)
Complete thermocyclingparameter	E	Yes	Materials and Methods: Quantitative real-time PCR.
Reaction setup (manual/robotic)	D	Yes	Manual setup.
Manufacturer of qPCR instruments	E	Yes	LightCycler 480; see also Materials and Methods: Quantitative real-time PCR.
qPCR VALIDATION			
Evidence of optimisation (from gradients)	D	Yes	Kits from Applied Biosystems; optimization guaranteed by the manufacturer.
Specificity (gel, sequence, melt, or digest)	E	Yes	Specificity guaranteed by manufacturer of the TaqMan assays.
For SYBR Green I, Cq of the NTC	E	Not applicable	-

Calibration curves with slope and Y-intercept	E	Yes	Material and Methods: Quantitative real-time PCR; see also Supporting Information, Methods S1.
PCR efficiency calculated from slope	E	Yes	Material and Methods: Quantitative real-time PCR; see also Supporting Information, Methods S1.
Confidence interval PCR efficiency or standard error	D	Yes	Supporting Information, Methods S1.
r ² of standard curve	E	No	Not provided by the LightCycler 480 software.
Linear dynamic range	E	Yes	Material and Methods: Quantitative real-time PCR; see also Supporting Information, Methods S1.
C _q variation at lowest concentration of the linear interval of calibration curves	E	Yes	Supporting Information, Methods S1.
Confidence intervals throughout range	D	No	-
Evidence for limit of detection	E	Yes	Material and Methods: Quantitative real-time PCR; see also Supporting Information, Methods S1. Measurements of all miRNAs were in linear dynamic range. Thus, it was not necessary to determine the LOD.
If multiplex, efficiency and LOD of each assay	E	N/A	-
DATA ANALYSIS			
qPCR analysis program (source, version)	E	Yes	Materials and Methods: Quantitative real-time PCR and Supporting Information, Methods S1.
C _q method determination	E	Yes	C _q >35 was decided as limit.
Outlier identification and disposition	E	Yes	There were no outliers.
Results of NTCs	E	Yes	NTC did not result in any amplification; C _q >40.
Justification of number and choice of reference genes	E	Yes	See Results/Discussion.
Description of normalization method	E	Yes	Materials and Methods: Data analysis, use of geNorm, NormFinder, BestKeeper.
Number and concordance of biological replicates	D	Yes	Figure 1 and 3: nonmalignant n = 17; low-grade tumor samples n = 20; high-grade tumor samples n = 21.
Number and stage (RT or qPCR) of technical replicates	E	Yes	Materials and Methods: Quantitative real-time PCR; triplicate measurements.
Repeatability (intra-assay variation, %CV)	E	Yes	Materials and Methods: Quantitative real-time PCR
Reproducibility (inter-assay variation, %CV)	D	Yes	Materials and Methods: Quantitative real-time PCR; in addition; biological replicates were preferred in favor of technical replicates.
Power analysis	D	No	-
Statistical methods for result significance	E	Yes	Materials and Methods: Data analysis; Results and Figure legends.
Software (source, version)	E	Yes	Materials and Methods: Data analysis.
C _q or raw data submission RDML	D	No	-

[&]E, essential information that must be submitted with the manuscript; D, desirable information that should be submitted with the manuscript if available.

Table S2 TaqMan assays for microRNAs and small nuclear and nucleolar RNAs

TaqMan assays from Applied Biosystems for the examined mature miRNAs and small nuclear RNAs are listed. The assay name, assay ID, miRBase accession number, miRBase ID, and the sequence are given for the miRNAs and correspondingly for the small nuclear and nucleolar RNAs. miRNAs were identified by the permanently assigned miRBase accession number, the miRBase-prescribed ID related to the miRBase version, and the sequence.

Assay name	Assay ID	miRBase accession no.	miRBase ID ^{&}	Sequence
hsa-miR-15a	000389	MIMAT0000068	hsa-miR-15a (v10.1) hsa-miR-15a-5p (v18)	UAGCAGCACAUAAUGGUUUUGUG
hsa-miR-20b	000580	MIMAT0001413	hsa-miR-20b (v10.1) hsa-miR-20b-5p (v18)	CAAAGUGCUCAUAGUGCAGGUAG
hsa-miR-29c	000587	MIMAT0000681	hsa-miR-29c (v10.1) hsa-miR-29c-3p(v18)	UAGCACCAUUUGAAAUCGGUUA
hsa-miR-101	002253	MIMAT0000099	hsa-miR-101 (v10.1) hsa-miR-101-3p (v18)	UACAGUACUGUGAUAACUGAA
hsa-miR-107	000443	MIMAT0000104	hsa-miR-107 (v10.1) hsa-miR-107 (v18)	AGCAGCAUUGUACAGGGCUAUC
hsa-miR-125a-5p	002198	MIMAT0000443	hsa-miR-125a-5p (v10.1) hsa-miR-125a-5p (v18)	UCCCUGAGACCCUUUAACCCUGUGA
hsa-miR-148b	000417	MIMAT0000759	hsa-miR-148b (v10.1) hsa-miR-148b-3p (v18)	UCAGUGCAUACACAGAACUUUUGU
hsa-miR-151-3p	002254	MIMAT0000757	hsa-miR-151-3p (v10.1) hsa-miR-151-3p (v18)	CUAGACUGAAGCUCCUUGAGG
hsa-miR-151-5p	002642	MIMAT0004697	hsa-miR-151-5p (v10.1) hsa-miR-151-5p (v18)	UCGAGGAGCUCACAGUCUAGU
hsa-miR-181a	000480	MIMAT0000256	hsa-miR-181a (v10.1) hsa-miR-181a-5p (v18)	AACAUUCAAACGCUGUCGGUGAGU
hsa-miR-181b	001098	MIMAT0000257	hsa-miR-181b (v10.1) hsa-miR-181b-5p (v18)	AACAUUCAUUGCUGUCGGUGGGU
hsa-miR-324-3p	002161	MIMAT0000762	hsa-miR-324-3p (v10.1) hsa-miR-324-3p (v18)	ACUGCCCCAGGUCUGCUGG

Assay name	Assay ID	miRBase accession no.	miRBase ID ^{&}	Sequence
hsa-miR-424	000604	MIMAT0001341	hsa-miR-424 (v10.1) hsa-miR-424-5p (v18)	CAGCAGCAAUUCAUUUUGAA
hsa-miR-513a-5p	002090	MIMAT0002877	hsa-miR-513a-5p (v10.1) hsa-miR-513a-5p (v18)	UUCACAGGGAGGUGUCAU
hsa-miR-874	002268	MIMAT0004911	hsa-miR-874 (v10.1) hsa-miR-874 (v18)	CUGCCCUGGCCCCGAGGGACCGA
hsa-miR-939	002182	MIMAT0004982	hsa-miR-939 (v10.1) hsa-miR-939 (v18)	UGGGGAGCUGAGGCUCUGGGGGUG

[&]miRNA ID in the miRBase version 10.1 and 18, respectively.

Gene name	Assay ID	Assay name	Sequence
RNU6B	001093	568915	CGCAAGGATGACACGCAAAATTCGTGAAGCGTCCATATTTT
RNU48	001006	568908	GATGACCCAGGTAACCTCTGAGTGTGTCGCTGATGCCATCACCCGACGCCCTCTGACC
Z30	001092	568917	TGGTATTGCCATTGCTTCACTGTTGGCTTTGACCAGGGTATGATCTCTTAATCTTCTCTGAGCTG

Table S3 List of the 101 miRNAs from the microarray for the identification of candidate reference miRNAs. All miRNAs of the microarray are listed that were flagged as "present" in all examined 24 tissue samples by the Genespring GX11 software.

	miRNA	Active sequence on the microarray
1	hsa-let-7a	AACTATACAACCTACTACCT
2	hsa-let-7b	AACCACACAACCTACTACC
3	hsa-let-7c	AACCATACAACCTACTACC
4	hsa-let-7d	AACTATGCAACCTACTACC
5	hsa-let-7e	AACTATACAACCTCCTACC
6	hsa-let-7f	AACTATACAATCTACTACCTC
7	hsa-let-7g	AACTGTACAAACTACTACCTC
8	hsa-let-7i	AACAGCACAAACTACTACCTC
9	hsa-miR-10b	CACAAATTCGGTTCTACAGGG
10	hsa-miR-15a	CACAAACCATTATGTGCTGCT
11	hsa-miR-15b	TGTAAACCATGATGTGCTGC
12	hsa-miR-16	CGCCAATATTTACGTGCTG
13	hsa-miR-17	CTACCTGCACTGTAAGC
14	hsa-miR-19a	TCAGTTTTGCATAGATTTGCA
15	hsa-miR-19b	TCAGTTTTGCATGGATTTGC
16	hsa-miR-20a	CTACCTGCACTATAAGCAC
17	hsa-miR-20b	CTACCTGCACTATGAGCAC
18	hsa-miR-21	TCAACATCAGTCTGATAAGC
19	hsa-miR-22	ACAGTTCTTCAACTGGCAG
20	hsa-miR-23a	GGAAATCCCTGGCAATGT
21	hsa-miR-23b	GGTAATCCCTGGCAATG
22	hsa-miR-24	CTGTTCTGCTGAACTGA
23	hsa-miR-25	TCAGACCGAGACAAGTGC
24	hsa-miR-26a	AGCCTATCCTGGATT
25	hsa-miR-26b	ACCTATCCTGAATTACTTGA
26	hsa-miR-27a	GCGGAACTTAGCCACTG
27	hsa-miR-27b	GCAGAACTTAGCCACTGT
28	hsa-miR-28-5p	CTCAATAGACTGTGAGCTCC
29	hsa-miR-29a	TAACCGATTTTCAGATGGTGC
30	hsa-miR-29b	AACACTGATTTCAAATGGTGC
31	hsa-miR-29c	TAACCGATTTCAAATGGTGCTA
32	hsa-miR-29c*	GAACACCAGGAGAAATCGGT
33	hsa-miR-30a	CTTCCAGTCGAGGATG
34	hsa-miR-30b	AGCTGAGTGTAGGATGTT
35	hsa-miR-30c	GCTGAGAGTGTAGGATGT
36	hsa-miR-30d	CTTCCAGTCGGGGA
37	hsa-miR-30e	CTTCCAGTCAAGGATGT
38	hsa-miR-34a	ACAACCAGCTAAGACACTGC
39	hsa-miR-92a	ACAGGCCGGGACAAGT
40	hsa-miR-93	CTACCTGCACGAACAG
41	hsa-miR-99b	CGCAAGGTCGGTTCTA
42	hsa-miR-101	TTCAGTTATCACAGTACTGT
43	hsa-miR-103	TCATAGCCCTGTACAATG
44	hsa-miR-106b	ATCTGCACTGTCAGCAC

	miRNA	Active sequence on the microarray
45	hsa-miR-107	TGATAGCCCTGTACAATGCT
46	hsa-miR-125a-3p	GGCTCCCAAGAACCTCA
47	hsa-miR-125a-5p	TCACAGGTTAAAGGGTCTC
48	hsa-miR-125b	TCACAAGTTAGGGTCTC
49	hsa-miR-126	CGCATTATTACTCACGGT
50	hsa-miR-130a	ATGCCCTTTTAAACATTGCA
51	hsa-miR-130b	ATGCCCTTTCATCATTGC
52	hsa-miR-140-3p	CCGTGGTTCTACCCCT
53	hsa-miR-140-5p	CTACCATAGGGTAAAACCACT
54	hsa-miR-141	CCATCTTTACCAGACAG
55	hsa-miR-142-3p	TCCATAAAGTAGGAAACACTACA
56	hsa-miR-143	GAGCTACAGTGCTTC
57	hsa-miR-145	AGGGATTCTGGGAAAAC
58	hsa-miR-148a	ACAAAGTTCTGTAGTGCCT
59	hsa-miR-148b	ACAAAGTTCTGTGATGCAC
60	hsa-miR-151-3p	CCTCAAGGAGCTTCAGT
61	hsa-miR-151-5p	ACTAGACTGTGAGCTCC
62	hsa-miR-181a	ACTCACCGACAGCGT
63	hsa-miR-181b	ACCCACCGACAGCA
64	hsa-miR-185	TCAGGAACTGCCTTTCT
65	hsa-miR-193b	AGCGGGACTTTGAGGG
66	hsa-miR-195	GCCAATATTTCTGTGCTGC
67	hsa-miR-199a-3p	TAACCAATGTGCAGACTACT
68	hsa-miR-199a-5p	GAACAGGTAGTCTGAACAC
69	hsa-miR-200a	ACATCGTTACCAGACAGT
70	hsa-miR-200b	TCATCATTACCAGGCAG
71	hsa-miR-200c	TCCATCATTACCCGG
72	hsa-miR-205	CAGACTCCGGTGGAAAT
73	hsa-miR-210	TCAGCCGCTGTCACAC
74	hsa-miR-223	TGGGGTATTTGACAAACTGAC
75	hsa-miR-320a	TCGCCCTCTCAAC
76	hsa-miR-324-3p	CCAGCAGCACCTGGGG
77	hsa-miR-331-3p	TTCTAGGATAGGCCAGGG
78	hsa-miR-342-3p	ACGGGTGCGATTTCTG
79	hsa-miR-361-5p	GTACCCCTGGAGATTC
80	hsa-miR-365	ATAAGGATTTTATAGGGGCATTA
81	hsa-miR-374a	CACTTATCAGGTTGTATTATAA
82	hsa-miR-378	CCTTCTGACTCCA
83	hsa-miR-423-5p	AAAGTCTCGCTCTCTG
84	hsa-miR-424	TTCAAAACATGAATTGCTGCTG
85	hsa-miR-425	TCAACGGGAGTGATCGTG
86	hsa-miR-429	ACGGTTTTACCAGACAGTA
87	hsa-miR-451	AACTCAGTAATGGTAACGGTTT
88	hsa-miR-483-5p	CTCCCTTCTTTCCCTC
89	hsa-miR-494	GAGGTTTCCCGTGTA
90	hsa-miR-497	ACAAACCACAGTGTGCTG
91	hsa-miR-513a-5p	ATGACACCTCCCTGTG

	miRNA	Active sequence on the microarray
92	hsa-miR-575	GCTCCTGTCCA ACTGGCT
93	hsa-miR-638	AGGCCGCCACCCGC
94	hsa-miR-768-3p_v11.0	GTCAGCAGTTTGAGTGTCAG
95	hsa-miR-768-5p_v11.0	ATCACTCCGTACTTTCATC
96	hsa-miR-801_v10.1	GTCGATTCCGCACGC
97	hsa-miR-874	TCGGTCCCTCGGG
98	hsa-miR-886-3p	AAGGGTCAGTAAGCACCCGC
99	hsa-miR-923_v12.0	AGTTTCTTTTCCTCCGC
100	hsa-miR-939	CACCCCCAGAGCC
101	hsa-miR-1225-5p	CCCCCACTGGG

Table S4 Spearman rank correlation coefficients (r_s) between the candidate reference genes. &

miR	miR-29	miR-101	miR-125a-5p	miR-148b	miR-151-3p	miR-151-5p	miR-181a	miR-181b	miR-324-3p	miR-424	miR-874	RNU6B	RNU48	Z30
miR-29c	r_s	0.575	0.470	0.393	0.272	0.500	0.201	-0.163	0.386	0.264	0.583	0.143	0.557	0.174
	p -value	0.0001	0.0001	0.002	0.039	0.0001	0.131	0.221	0.003	0.045	0.0001	0.283	0.0001	0.192
miR-101	r_s	0.575	0.621	0.551	0.144	0.622	0.105	0.028	0.630	0.249	0.489	-0.074	0.196	0.257
	p -value	0.0001	0.0001	0.0001	0.280	0.0001	0.434	0.834	0.0001	0.059	0.0001	0.581	0.141	0.051
miR-125a-5p	r_s	0.470	0.621	0.380	0.266	0.787	0.314	0.150	0.709	0.306	0.341	-0.150	0.078	0.186
	p -value	0.0001	0.0001	0.003	0.044	0.0001	0.016	0.263	0.0001	0.020	0.009	0.261	0.560	0.161
miR-148b	r_s	0.393	0.551	0.380	0.649	0.530	0.357	0.220	0.440	0.315	0.368	0.137	0.381	0.524
	p -value	0.002	0.0001	0.003	0.0001	0.0001	0.006	0.096	0.001	0.016	0.004	0.306	0.003	0.0001
miR-151-3p	r_s	0.272	0.144	0.266	0.649	0.503	0.304	0.245	0.135	0.149	0.374	0.216	0.433	0.489
	p -value	0.039	0.280	0.044	0.0001	0.0001	0.021	0.064	0.313	0.266	0.004	0.104	0.001	0.0001
miR-151-5p	r_s	0.500	0.622	0.787	0.503	0.503	0.229	0.089	0.672	0.173	0.349	-0.115	0.239	0.298
	p -value	0.0001	0.0001	0.0001	0.0001	0.0001	0.084	0.505	0.0001	0.194	0.007	0.390	0.071	0.023
miR-181a	r_s	0.201	0.105	0.314	0.304	0.229	0.746	0.746	0.200	0.586	0.147	0.048	0.355	0.289
	p -value	0.131	0.434	0.016	0.006	0.021	0.0001	0.0001	0.132	0.0001	0.272	0.722	0.006	0.028
miR-181b	r_s	-0.163	0.028	0.150	0.220	0.245	0.746	0.746	0.091	0.350	-0.046	0.053	0.052	0.231
	p -value	0.221	0.834	0.263	0.096	0.064	0.0001	0.0001	0.498	0.007	0.733	0.694	0.698	0.081
miR-324-3p	r_s	0.386	0.630	0.709	0.440	0.135	0.200	0.091	0.672	0.214	0.224	-0.275	-0.045	0.092
	p -value	0.003	0.0001	0.0001	0.001	0.313	0.132	0.498	0.107	0.090	0.090	0.037	0.736	0.491
miR-424	r_s	0.264	0.249	0.306	0.149	0.173	0.586	0.350	0.214	0.151	0.151	0.184	0.352	0.256
	p -value	0.045	0.059	0.020	0.266	0.194	0.0001	0.007	0.107	0.257	0.166	0.166	0.007	0.052
miR-874	r_s	0.583	0.489	0.341	0.374	0.349	0.147	-0.046	0.224	0.151	0.200	0.504	0.504	0.012
	p -value	0.0001	0.0001	0.009	0.004	0.007	0.272	0.733	0.090	0.257	0.132	0.0001	0.0001	0.930
RNU6B	r_s	0.143	-0.074	-0.150	0.137	-0.115	0.048	0.053	-0.275	0.184	0.200	0.436	0.436	0.197
	p -value	0.283	0.581	0.261	0.306	0.104	0.722	0.694	0.037	0.166	0.132	0.001	0.001	0.139
RNU48	r_s	0.557	0.196	0.078	0.381	0.433	0.355	0.052	-0.045	0.352	0.504	0.436	0.353	0.353
	p -value	0.0001	0.141	0.560	0.003	0.001	0.006	0.698	0.736	0.007	0.0001	0.001	0.007	0.007
Z30	r_s	0.174	0.257	0.186	0.524	0.489	0.289	0.231	0.092	0.256	0.012	0.197	0.353	0.353
	p -value	0.192	0.051	0.161	0.0001	0.0001	0.028	0.081	0.491	0.052	0.930	0.139	0.007	0.007

&Correlations marked in yellow indicate r_s values ≥ 0.6 .

Methods S1 qPCR validation experiments according to the MIQE guidelines with respect to the calibration curves and the dynamic range of measurements. Calibration curves were generated with diluted cDNAs (at least six dilution steps). The Cq values were calculated automatically by the LightCycler software, release 1.5.0 using the "second derivative maximum" method. The slopes, intercepts, and errors of the regression lines of the calibration curves from these dilution series and the PCR efficiencies ($E=10^{-1/\text{slope}}$) including the dynamic range and the Cq variation at the lower limit (the endpoint of the linear dynamic range) were also calculated by this software. Data and curves of the miR-26a, miR-151-3p, and RNU48 are exemplarily shown as follows. Since efficiencies did only differ in the second decimal place confirming the manufacturer's information that the TaqMan miRNA assays run with equivalent amplification efficiencies, we used the data of the calibration curve of miR-26a for all miRNAs and those of the curve of RNU48 for all RNUs.

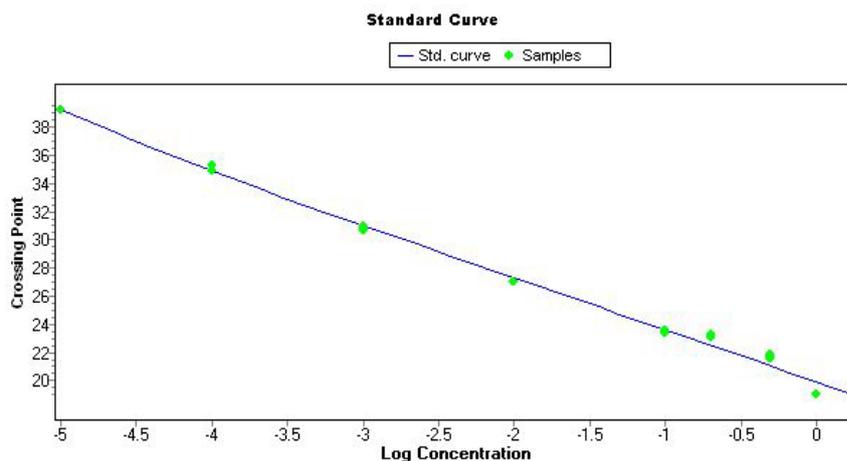
Gene	PCR-Efficiency	Slope	Y-Intercept	Error ^{&}	Linear dynamic range [#]	Cq variation at the lowest limit (SD) [§]
miR-26a	1.862	-3.703	19.82	0.0459	19.02-35.29	0.17
miR-151-3p	1.884	-3.636	25.30	0.0653	24.31-35.60	0.11
RNU48	1.813	-3.870	22.22	0.0339	20.84-33.76	0.77

[&] The error value is the mean squared error of the single data points fit to the regression line, according to the definition given in the handbook of the LightCycler software.

[#] The linear dynamic range represents the range of the Cq values between the highest and the lowest concentration of linear interval of the calibration curve.

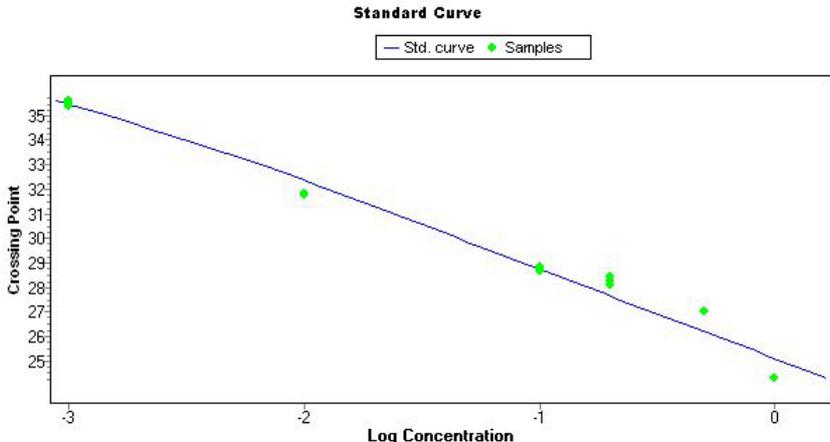
[§] Cq variation given as SD at the endpoint of the linear dynamic range that corresponds to the lowest concentration in the linear interval of the calibration curve.

Calibration curve of hsa-miR-26a



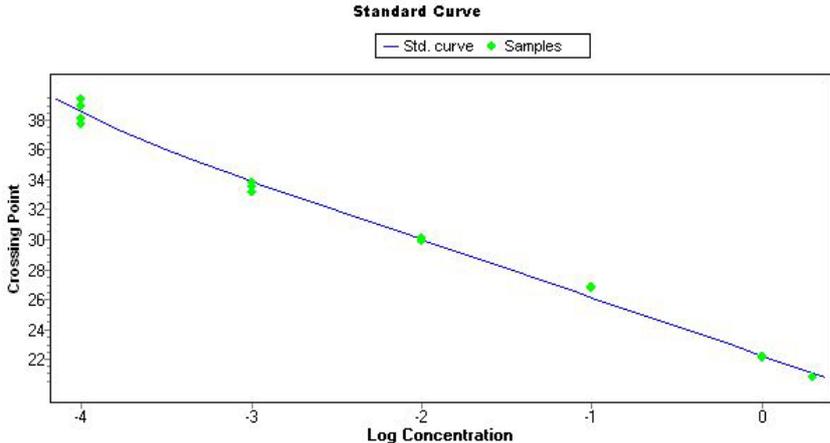
Error: 0.0459
 Efficiency: 1.862
 Slope: -3.703
 Y-Intercept: 19.82

Calibration curve of hsa-miR-151-3p



Error: 0.0653
Efficiency: 1.884
Slope: -3.636
Y-Intercept: 25.30

Calibration curve of RNU48



Error: 0.0339
Efficiency: 1.813
Slope: -3.870
Y-Intercept: 22.22

Lebenslauf

Mein Lebenslauf wird aus datenschutzrechtlichen Gründen in der elektronischen Version meiner Arbeit nicht veröffentlicht.

Lebenslauf

Mein Lebenslauf wird aus datenschutzrechtlichen Gründen in der elektronischen Version meiner Arbeit nicht veröffentlicht.

Publikationsliste

1. **Ratert N**, Meyer HA, Jung M, Mollenkopf HJ, Wagner I, Miller K, Kilic E, Erbersdobler E, Weikert S, Jung K. Reference miRNAs for miRNAome analysis of urothelial carcinomas. PloS One 2012;7:e39309.

Impact Factor: 4.092

2. **Ratert N**, Meyer HA, Jung M, Lioudmer P, Mollenkopf HJ, Wagner I, Miller K, Kilic E, Erbersdobler, Weikert S, Jung K. MicroRNA profiling identifies candidate miRNAs for bladder cancer diagnosis and clinical outcome. J Mol Diagn 2013; accepted May 13, 2013.

Impact Factor: 3.576

Postervorträge

1. **Ratert N**, Wotschofsky Z, Fendler A, Weikert S, Busch J, Miller K, Meyer HA, Stephan C, Jung M, Erbersdobler A, Mollenkopf HJ, Jung K. Validierte miRNA-Referenzgene bei urologischen Tumoren. [Abstract P1.1: Urologe 2012;51:99-100].

Impact Factor: 0.497

2. **Ratert N**, Lioudmer P, Meyer HA, Jung M, Mollenkopf HJ, Wagner I, Miller K, Kilic E, Erbersdobler A, Weikert S, Jung K. MicroRNA profiling in bladder cancer identifies miR-141 and miR-205 as prognostic factors. [Abstract P2.3: Urologe 2013;52:112].

Impact Factor: 0.497

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