Concentration of circulating miRNA-containing particles in serum enhances miRNA detection and reflects CRC tissuerelated deregulations

SUPPLEMENTARY MATERIALS AND METHODS

RNA quality checks

To ensure sample quality and quantity, different quality controls were undertaken. To detect the presence of inhibitors, and as a technical control, a UniSp6 RNA spike-in probe (Exiqon, Vedbaek, Denmark) was included in the reverse transcription reaction (RT) to evaluate RT performance variation in all samples. A negative (water) control was co-profiled with the samples to exclude any RNA contamination. Moreover, a sample hemolysis check was performed using the relative expression of two hemolysis indicators, the erythrocyte-specific miRNA-451 and the stable miRNA-23a [1]. Samples showing a hemolysis index below 7.0, calculated using the delta quantification cycle of both miRNAs $[dC_q(miR-23a - miR-451)$ ratio], were considered as of good quality, i.e. no or very minimal signs of hemolysis.

cDNA synthesis and qPCR

The cDNA amplification was performed in a LightCycler® 480 Real-Time PCR System (Roche) in 384 well plates and analyzed using the Roche LC software. The applied qPCR method in the current study has been recently shown to possess a decent efficiency [2].

Data filtering and analysis

The amplification efficiencies in the real-time PCR experiments were determined by analyzing the amplification curves using algorithms similar to the LinReg software package [3]. The amplification efficiencies in the real-time PCR experiments ranged between 1.8 and 2.1. Individual reactions with efficiencies <1.6 were excluded from the dataset. All assays were examined for distinct melting curves and the Tm's were inspected to be within known specifications for each particular experiment. Any sample crossing point (C_q) value, which was not detected with at least 5 C_q cycles less than the corresponding negative control, and with a C_q < 37 (upper limit), was considered undetectable. Data that did not pass the quality control criteria were omitted from further analysis.

It is well known that normalization of qPCR miRNA expression data is a persistent challenge and choosing the proper strategy is crucial to obtain reliable results. In the current study, two data normalization methods [4, 5] were independently applied to the same raw data resulting from the initial screen. First, the global mean normalization method, which is based on the average of the assays detected in all samples [4], was applied. This method has been shown to be suitable to normalize qPCR miRNA expression data across different cell types and tissues [6]. The stability of the average of the selected miRNAs for normalization was higher than any single microRNA in the data set as confirmed by the Normfinder software [7]. To further improve our miRNA-selection, the modified global mean normalization strategy, based on the attribution of equal weight to each individual miRNA during normalization [5], was independently applied to the same raw data resulting from the initial screen (Figure 1). Multiple testing correction according to the Benjamini-Hochberg method [8] was applied. P-values <0.05 were considered statistically significant.

To assess miRNA expression differences between samples, fold changes were calculated and t-tests and multiple testing correction according to the Benjamini-Hochberg method were applied. In addition, heat map and unsupervised hierarchical clustering were carried out using the Euclidian distance. ANOVA analysis was also performed to display the expression changes of different miRNAs across all tested samples in five groups. The RNA/cDNA quality controls, miRNA qPCR profiling experiments as well as the statistical analyses for this study were performed at Exiqon services (Vedbaek, Denmark).

Furthermore, the miEAA tool (http://www.ccb.unisaarland.de/mieaa_tool) was used to perform different overrepresentation and enrichment analyses of the identified 22 miRNAs in the CRC's particle-concentrated serum fractions in different cellular pathways and reported relations to diseases or affected organs/tissues.

SUPPLEMENTARY RESULTS

Initial screen: Expression differences between whole sera and particle-concentrated sera

The principal component analysis showed a separation between the matched particle-concentrated and particle-depleted (supernatant) fractions (Supplementary Figure S2). This was further supported by heat map and unsupervised hierarchical clustering analyses (Supplementary Figure S3).

Each of the two normalizations approaches yielded 44 statistically significant candidates, which showed differential expression between matched whole sera and particle-concentrated fractions. Both methods yielded 43 of 44 of overlapping miRNAs (Supplementary Table S1). For further analysis, the 43 overlapping miRNAs plus the two uniquely identified miRNAs (miR-143; miR-423) in both normalized datasets were considered in the validation experiments.

Tissue preservation conditions can influence miRNA expression patterns

To assess the effect of different RNA preservation methods for tissue miRNA detection, six of the tested 25 CRC tissue specimens were analyzed as paired samples from the same tissues stored either after snap-freezing in liquid nitrogen or after fixation in RNA-later reagent. A pronounced impact of the preservation method was observed based on the number of the detected miRNAs and on the average C_q values measured (Supplementary Table S5; Supplementary Figure S4). Using a paired t-test, 13 miRNAs were found to be differentially expressed (Supplementary Figure S5). Differential expression of two miRNAs remained significant after correction for multiple testing (Supplementary Table S6). While miR-101-3p was mainly upregulated in tissue samples preserved in liquid nitrogen, miR-15b-5p showed upregulation primarily in tissue samples stored in RNA-later. The reason for this differential expression or the slightly better performance of tissue samples preserved in liquid nitrogen remains unclear. Yet, our results are in agreement with other studies that underscore the general suitability of the routinely processed tissue samples for miRNA detection. Differential expression levels of miRNAs, caused either by different tissue storage strategies [9] or storage time [10] have been reported in similar studies. These observations should be taken into account when interpreting expression data and comparing them with those of other studies. Together with other findings, we considered only the data of the tissues preserved in liquid nitrogen (n=15) in our downstream analyses, in order to avoid additional confounding factors.

miRNA expression in particle-concentrated serum mirrors deregulation of tumor tissues

Seven of the 22 miRNAs in Table 1 showed inconsistent expression between the particle-concentrated serum fractions and the corresponding tissue samples. This may provide some mechanistic insights on the miRNA release into the extracellular environment by cancer cells. In brief, the expression of five of these seven miRNAs (miR-22-5p, -223-3p, -320b, -335-5p, and -144-3p) were significantly upregulated in the particle-concentrated sera of patients compared to those of the respective controls. The first three of these five miRNAs (miR-22-5p,-223- 3p and -320b) did not show any significant deregulation when tissue samples were compared to controls' particleconcentrated fractions. But when the level of these three miRNAs in the patients' particle-concentrated fractions was compared to that of the corresponding tissue samples, a significant downregulation was observed in the tissue samples (P-values, 0.0001, 0.0081 and 0.0030, respectively). The other two miRNAs, miR-335-5p, and -144-3p, were even more downregulated in tissue samples than in the controls' particle-concentrated serum fractions (P-values, 0.0024 and 0.0001, respectively) and in patients' particle-concentrated serum fractions (P-values, 2.48×10^{-9} and 6.45×10^{-7} , respectively). The expression of the remaining two miRNAs, let-7d-3p and miR-342-3p, were significantly downregulated in patients' particle-concentrated serum fractions compared to the respective controls (P-values, 0.0007 and 0.0064, respectively), but showed significant upregulation in tissue samples compared to the matched patients' particleconcentrated serum fractions (P-values, 0.0049 and 0.0003, respectively) (Table 1).

miRNA *in silico* **enrichment analyses and relation to cancer and inflammatory pathways**

Significant over representation of the identified 22 miRNAs in the particle-concentrated serum fractions have been reported in the context of many diseases, such as "carcinoma" and "adenocarcinoma" (22 and 17 miRNAs; P-values 0.0002 and 0.0002, respectively), "neoplasm_metastasis" (21 miRNAs; P-value, 1.12×10^{-12} 05), "inflammation" (16 miRNAs; P- value 0.0002), and "colonic_neoplasms" (10 miRNAs; P-value, 0.0117). "Organs miRWalk" analysis using the miEAA tool showed significant over representation of our 22 miRNA candidates in different organs or tissues, for instance in "blood serum" (20 miRNAs; P-value, 1.6×10^{-6}) and "colon" (15 miRNAs; P-value 0.0002). Moreover, "pathway miRWalk" analysis by the miEAA tool resulted in significant enrichment of most of these 22 miRNAs in inflammation- and cancer-related pathways, such as "apoptosis" (22 miRNAs; P-value, 0.0002), "mTOR" (19 miRNAs; P-value, 0.0002), "RAS" (19 miRNAs; P-value 0.0002), "Inflammation" (20 miRNAs; P-value 0.0002), "TLR" (21 miRNAs; P-value 0.0005), "IL-6" (18 miRNAs; P-value, 0.0006), "STAT3" (14 miRNAs; P-value, 0.0006), "cell cycle" (22 miRNAs; P-value, 0.0009), "P53" (21 miRNAs; P-value, 0.0009), "NF-κB"

(16 miRNAs; P- value, 0.0012), "cytokine-cytokinereceptor-interaction" (21 miRNAs; P-value, 0.0014), "CRC" (22 miRNAs; P-value, 0.0019), "WNT" (21 miRNAs; P-value, 0.0026), and "Jak-STAT-signaling" (20 miRNAs; P-value, 0.0062) (other details in Supplementary Table S8).

SUPPLEMENTARY DISCUSSION

Concentration of miRNA-carrying particles improves detection of CRC-related miRNAs

Our observation of a lower detection rate of miRNA in the unprocessed serum compared to the concentrated miRNA-containing particles in corresponding serum seems reasonable based on recent observations and findings. Chevillet and colleagues [11] suggested that miRNA pattern after enrichment of potentially miRNArich subpopulations of exosomes may be different from the overall miRNA signature in the same biofluid without enrichment. Moreover, Cheng and colleagues [12] emphasized that the presence of miRNA in cell-free samples is diluted in the plasma and serum volume, and this low abundance can be regarded as a confounding factor in the detection of miRNA by qPCR. Altogether, these observations support the applied strategy in our study, of enriching the "bulk" of miRNA-containing circulating particles (extracellular vesicles and others structures such as RNA-protein and RNA-lipoprotein complexes) as a reasonable approach in the situation of miRNA biomarker discovery. This may also plausibly explain the higher detection rate of miRNA candidates in the concentrated serum fraction compared to the corresponding unprocessed serum.

The number of differentially regulated miRNAs (28 out of 742; \sim 4%; Figure 2) should be better put into context of other studies. In the present study, following high sensitivity qRT-PCR detection, we applied unique filtering and stringent selection criteria. This strategy resulted in a reliable set of miRNA candidates and is comparable in terms of numbers of regulated transcripts to other studies. For example, the results of a microarraybased analysis followed by a qRT-PCR validation [13], ended up with seven miRNAs that showed significantly higher expression levels in serum EVs of CRC patients $(n=88)$ than in healthy controls $(n=11)$, as well as downregulation after surgical resection of tumors. Similarly, these seven miRNAs had higher expression levels in colon cancer cell lines (n=5) than in a normal colonderived cell line. Notably, three of these seven miRNAs, namely miR-21, miR-23a and miR-223, also showed significantly higher expression in our particle-concentrated serum fractions (Table 1). In another study [14], out of 1547 miRNAs profiled using a Universal RT microRNA PCR system (GeneCopoeia, USA) in two pooled tissue

samples (28 tumors and 28 paired normal controls), only 93 miRNAs (6%) were firstly identified with significantly dysregulation Colorectal adenocarcinoma relative to normal tissues. Further filtering and downstream analyses ended up with only 32 miRNAs (2.1%) that were able to distinguish cancer tissues from normal tissues, as well as to identify well- and moderately differentiated cancers. Remarkably, a realistic subset of these 32 miRNAs, for example miR-23b, miR-125b, miR-26a, let-7i, miR-99a, and miR-100, were also highlighted in our study as potential biomarker candidates (Figures 2 and 4). In a third study [12], which utilized a different EVs isolation kit system (Norgen Biotek, Canada) followed by nextgeneration sequencing, 15 and 40 miRNAs were detected (out of 386 and 412 miRNAs; 4% and 10%, respectively) to be confined inside serum- and plasma-EVs, respectively. Those 15 and 40 miRNAs were not detectable in whole (PAXgene) blood. Similarly, in our study 15 of 28 (>50%) miRNAs were found to be only deregulated in patients' particle-concentrated sera (Figure 2).

Concentration of miRNA-carrying particles improved detection of CRC-related signatures

Examples of miRNAs with consistent upregulation in particle-concentrated CRC sera and tissues, but not in whole sera, are miR-125 and miR-23a. While miR-125 is known as an important prognostic indicator for CRC [15], miR-23a is a critical regulator of CRC migration and metastasis [16].

SUPPLEMENTARY FOOTNOTE TO TABLE 1

In each comparison presented in Table 1, two respective columns are shown, one for the fold changes and the other for the adjusted/corrected P-values. When the fold change has a positive value, this means that the first item in the respective comparison showed higher miRNA expression level than the second item of the same comparison; and when the P-value does not indicate significance, implies that the compared expression levels are statistically the same (i.e., similar expression).

In the first (I) and third (III) comparison, it is clear that the first 10 miRNAs showed higher expression (fold changes with positive values and significant P-values \leq 0.05) in the particle-concentrated CRC serum samples and tumor tissues compared to particleconcentrated sera of the controls, respectively. Yet, in the second comparison (II. Particle-Concentrated Sera (CRC) vs. Tissue (CRC)), seven miRNAs are described by negative fold changes (printed in italics in Table 1), which reflect lower expression levels of these miRNAs in particle-concentrated CRC sera than in CRC tissue samples. The corresponding P-values of the remaining three miRNAs, although having positive fold changes,

indicated non-significance, hence their expression levels were statistically not different. Based on this evaluation, we concluded that the first 10 miRNAs in Table 1 showed consistent expression levels in sera and tissues of the investigated CRC cases.

The same holds true for the five down-regulated miRNAs (miR-486-5p, -93-5p, -92a-3p, -146a-5p, and -221-3p). Here, in the first (I) and third (III) comparison, it is clear that these 5 miRNAs showed lower expression in the particle-concentrated CRC serum samples and CRC tissues samples compared to the particle-concentrated serum samples of the controls (fold changes with negative values and P-values ≤ 0.05). Yet, in the second comparison, two of these miRNAs with barely negative fold change showed P-values not reaching significance. Therefore, these expression levels (of miR-92a-3p, -221- 3p) are rated as equal. This was also concluded for miR-93-5p and miR-146a-5p, which also had non-significant (positive) fold-changes. But remarkably, miR-486-5p showed the highest downregulation in the CRC tissues compared to the corresponding sera with high significance $(P-value, 1.14\times10^{-14})$.

Accordingly, a total of 15 out of the 22 miRNA candidates (68%) showed consistent expression patterns in the sera and tissues of CRC samples. Moreover, the ANOVA results, across all the tested samples (Supplementary Table S7), clearly support the aforementioned interpretation. Taking miR-21-5p as an example, the average normalized C_q values (dC_q) were higher in the CRC tissues than in the particle-concentrated CRC sera (0.8377 vs. -0.28552), which point to higher expression in the tumor than in the corresponding particle-concentrated sera, which is in agreement with our interpretation of the results presented in Table 1.

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SUPPLEMENTARY FIGURES AND TABLES

Supplementary Figure S1: Initial screen – The number of the detected miRNAs in the tested three serum compartments (whole [b], particle-concentrated [a], and particle-depleted[s]). This graphical illustration shows the number of miRNA detected in the initial screen, using qPCR of 742 miRNA panel, in matched samples (whole serum, particle-concentrated sera, and particle-depleted sera (supernatant)), of 10 CRC patients with metastasis (all at tumor stage IV). More miRNAs were detected in particle-concentrated fractions than in corresponding unprocessed serum and particle-depleted samples. The observation that a considerable number of detected miRNAs were picked up in the particle-depleted serum samples may imply that the applied polymer-based precipitation method needs further improvements in order to enable better concentration of different miRNA forms and miRNA-containing particles in serum samples. Moreover, although these results clearly underscore higher detection rate of circulating miRNAs in the particle-concentrated serum fractions, we can't rule out the influence of the larger volume of the sera used in the concentration step compared to the unprocessed serum. Moreover, qPCR data may be influenced by a threshold cDNA input. To deal with these confounding factors, we applied two different data normalization methods and didn't pay so much attention to the differences in the number of detected miRNAs in the initial screening stage. Rather, in the downstream analyses, after the validation experiments, we focused on comparing the differentially deregulated miRNAs in matched serum fractions of the patient and respective healthy control samples (Figures 2 and 3) and monitoring their expression in the paired tissue samples (Table 1 and Supplementary Tables S7 and S10).

Supplementary Figure S2: Initial screen – Principal component analysis (PCA) plot. The principal component analysis is performed on all samples, and on the 30 miRNAs most differentially expressed. The normalized (dC_q) values have been used for the analysis. The sample source separates relatively well on the first primary component but they also seem to somewhat pair on the second component (green = particle-concentrated serum samples; blue = whole serum and red= supernatant or particle-depleted serum samples). Here, the particle-concentrated serum samples (green boxes) are clearly separated from the particle-depleted serum (supernatant) samples (red boxes) and the whole sera (blue boxes) in the middle.

Supplementary Figure S3: Initial screen – Heat map and unsupervised hierarchical clustering of the samples tested. The heat map diagram shows the result of the two-way hierarchical clustering of miRNAs and samples. Each row represents one miRNA and each column represents one sample. The miRNA clustering tree is shown on the left. The color scale shown at the top illustrates the relative expression level of a miRNA across all samples: red color represents an expression level above mean, blue color represents expression lower than the mean and grey color represents missing data. The clustering is performed on all samples, and on the 30 most differentially expressed miRNAs (note that missing data are more common in the supernatant samples). The normalized (dC_q) values have been used for the analysis. A clear differential expression of the tested miRNAs emerged among different serum fractions investigated. (Serum exosomes in this figure refers to particle-concentrated serum samples).

Supplementary Figure S4: miRNAs detected in all tested CRC tissue samples (paired, n=25). On average, 42 microRNA were detected per sample. A general high quality and quantity of RNAs were isolated from all tissue samples. Expression profiling of the selected 45 miRNAs, using the custom qPCR panel, showed a general good performance of all the 25 tissue samples tested as shown on this Figure: an average of 42 of the tested 45 miRNAs were detected in all tested tissue samples.

Supplementary Figure S5: Differentially expressed miRNAs across CRC tissue samples. When comparing the two tissue storage conditions (snap-frozen in liquid nitrogen or RNA-later reagent), using a paired t-test, 13 miRNAs were found to differentially expressed using a cutoff of P-value < 0.05.

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(Particle Size/Concentration)

Supplementary Figure S6: Comparison of particle isolation methods using Nanoparticle Tracking Analysis (NTA). miRNA-containing particles were isolated (from serum free cancer cell culture supernatant; 1:10 dilution) by three different methods, namely (I) ExoQuick precipitation (SBI/BioCat, Heidelberg, Germany), (II) CellGS Exospin (Cell Guidance Systems, Cambridge, UK), and (III) the ultracentrifugation procedure. The results of NTA measurement (NanoSight NS300; software version 2.3) were compared regarding the particle size. The mean sizes in all three measurements were 200 nm, 222 nm and 223 nm, respectively. The concentration of miRNA-carrying particles obtained using ultracentrifugation protocol was higher, than that of the kit-base systems (ExoQuick and ExoSpin). This might be due to the higher volume (twofold of the input material) used in the ultracentrifugation experiments to get a comparable visible particle-precipitated pellet.

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(Particle Size/Concentration)

Supplementary Figure S7: Effect of storage conditions on isolated particle's size, concentration and reproducibility of NTA analysis. Two particle-concentrated serum samples (1:250 dilution) were enriched from patients' sera by ExoQuick precipitation. While isolated particles from one sample A (I) were kept at 4°C, particles from sample B (II) were stored at -80°C. NTA analyses (NanoSight NS300; software version 2.3) shows that particle size mean was 145 nm in both cases (I and II). A slight difference in particle concentration was noticed. To check for the reproducibility of the NTA measurement, each sample was measured at two different time points. As shown in the lower Figure III, the NTA measurement was reproducible in both samples A and B. It is noticeable that, although in each independent analysis (shown in Supplementary Figure S6 and S7) a peak of particle size close to 100-150 nm has been observed, differences in mean particle sizes (and heterogeneous peak shoulders) have been observed in different preparations, even using the same particle isolation method. This can be affected by different factors, such as different sample sources (serum free cell culture supernatant vs. serum samples, in the analyses shown in Supplementary Figures S6 and S7, respectively), sample dilutions (1:250 vs. 1:10, respectively), ratio of protein aggregates in the preparation, and respective heterogeneity of (co)isolated particles from different samples.

Supplementary Table S1: Expression differences and fold changes of detected miRNAs of matched whole serum and particle-concentrated serum samples of CRC patients (initial screen).

See Supplementary File 1

Supplementary Table S2: Samples tested.

See Supplementary File 2

Supplementary Table S3: Differentially expressed miRNAs between particle-concentrated serum of CRC and control samples

miRNAs in bold prints (n=12) represent those (overlapping) miRNAs that were also detected to be differentially expressed in the whole serum CRC samples (see Supplementary Table S4 and Figure 2). The remaining miRNAs (n=10) are the ones that were detected to be differentially expressed in particle-concentrated CRC samples only (and not in the matched whole serum CRC samples). Positive ddC_q values indicate upregulation and negative values indicate downregulation.

miRNAs in bold prints (n=12) represent those (overlapping) miRNAs that were also detected to be differentially expressed in particle-concentrated samples (see Supplementary Table S3 and Figure 2).

Supplementary Table S5: Difference in miRNA content and qPCR average C^q values between CRC tissue samples (snap-frozen and stored liquid nitrogen) to the matched samples (stored in RNA-later reagent (n=6))

	Average (snap-frozen liquid nitrogen)	Average (RNA-later reagent)	StDev	P-value
miRNAs detected	44.3	42.2	1.35	0.0061
Average C_{a}	25.26	27.25	1.46	0.0147

The above table lists the average number of assays detected in the CRC tissue samples provided in replicates either snapfrozen in liquid N2 or stored in RNA-later. As shown in this Table, more miRNAs were detected in tissue samples stored in snap-frozen liquid nitrogen than those stored in RNA-later (average 44.3 vs. 42.2 miRNAs, respectively; standard deviation: 1.35, P-value of the t-test: 0.0061). The same applies to the C_q values of the qPCR reactions, which were 25.26 and 27.25 for samples stored in liquid nitrogen and RNA-later, respectively (standard deviation 1.46; P value 0.0147). Both t-tests thus indicated that there were significant differences between the number and C_q values of the expressed miRNAs in tissue samples stored in different conditions.

Supplementary Table S6: Differentially expressed miRNAs between matched CRC tissue samples, snap-frozen in liquid nitrogen (n=6) or stored in RNA-later reagent (n=6)

Table showing microRNA names, average dC_q values, followed by ddC_q and p-values from a paired test between the two groups.

Shown in this Table the top 10 miRNAs that were differentially expressed between matched tissue samples sanp-frozen in liquid nitrogen and RNA-later reagent.

Two miRNAs, namely, miR-101-3p qand miR-15b-5p, remained with significantly P-values (marked in bold) after multiple testing corrections.

Supplementary Table S7: The results of the ANOVA analysis across all the tested samples.

See Supplementary File 3

Supplementary Table S8: Selected results from the miEAA analysis tool of the identified 22 miRNAs in the particleconcentrated CRC Sera.

See Supplementary File 4

Supplementary Table S9: Possible interplay of the identified 22 miRNAs in the particle-concentrated sera with key inflammatory and cancer-related pathways.

See Supplementary File 5

Supplementary Table S10: Particle-concentrated serum miRNA patterns of CRC patients with metastasis reflect tissuerelated expression.

See Supplementary File 6

Supplementary Table S11: Possible interplay of the identified 22 miRNAs in the particle-concentrated sera of metastatic CRC patients with key inflammatory and cancer-related pathways.

See Supplementary File 7