Profilig ofmetastatic small intestine neuroendocrine tumors reveals characteristic miRNAs detectable in plasma

Michaela Bowden1, Chensheng W. Zhou1, Sui Zhang1, Lauren Brais1, Ashley Rossi1, Laurent Naudin2, Arunthi Thiagalingam3, Ewa Sicinska1 and Matthew H. Kulke1

1Department of Medical Oncology, Dana-Farber Cancer Institute, Boston, MA, USA
2IPSEN Innovation, Les Ulis, France
3IPSEN Bioscience Inc., Cambridge, MA, USA

Correspondence to: Michaela Bowden, email: michaela_bowden@dfci.harvard.edu

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ABSTRACT

Background: Current diagnostic and prognostic blood-based biomarkers for neuroendocrine tumors are limited. MiRNAs have tumor-specific expression patterns, are relatively stable, and can be measured in patient blood specimens. We performed a multi-stage study to identify and validate characteristic circulating miRNAs in patients with metastatic small intestine neuroendocrine tumors, and to assess associations between miRNA levels and survival.

Methods: Using a 742-miRNA panel, we identified candidate miRNAs similarly expressed in 19 small intestine neuroendocrine tumors and matched plasma samples. We refined our panel in an independent cohort of plasma samples from 40 patients with metastatic small intestine NET and 40 controls, and then validated this panel in a second, large cohort of 120 patients with metastatic small intestine NET and 120 independent controls.

Results: miRNA profiling of 19 matched small intestine neuroendocrine tumors and matched plasma samples revealed 31 candidate miRNAs similarly expressed in both tissue and plasma. We evaluated expression of these 31 candidate miRNAs in 40 independent cases and 40 normal controls, and identified 4 miRNAs (miR-21-5p, miR-22-3p, miR-29b-3p, and miR-150-5p) that were differently expressed in cases and controls (p<0.05). We validated these 4 miRNAs in a separate, larger panel of 120 cases and 120 controls. We confirmed that high circulating levels of miR-22-3p (p<0.0001), high levels of miR 21-5p, and low levels of miR-150-5p (p=0.027) were associated with the presence of metastatic small intestine NET. While levels of 29b-3p were lower in cases than in controls in both the initial cohort and the validation cohort, the difference in the validation cohort did not reach statistical significance. We further found that high levels of circulating miR-22-5p, high levels of circulating miR-22-3p and low levels of circulating miR-150-5p were each independently associated with shorter overall survival. A combined analysis using all three markers was highly prognostic for survival (HR 0.47, 95% CI 0.27-0.82).

Conclusions: Our study suggests that elevated circulating levels of miR-21-5p and miR-22-3p and low levels of miR-150-5p are characteristic in patients with metastatic small intestine neuroendocrine tumors, and further suggests that levels of these miRNAs are associated with overall survival. These observations provide the basis for further validation studies, as well as studies to assess the biological function of these miRNAs in small intestine neuroendocrine tumors.
INTRODUCTION

Accurate, blood-based biomarkers to facilitate diagnosis and to monitor neuroendocrine tumor response to treatment are currently limited. Chromogranin A, a protein secreted by neuroendocrine cells, is a commonly used biomarker in patients with neuroendocrine tumors and can be obtained at diagnosis or in follow up of patients with resected or metastatic disease [1]. However, chromogranin A has limited sensitivity and specificity in this setting [2-5]. Measurement of plasma or urinary 5-HIAA, a serotonin metabolite is also used as a clinical tool in the care of patients with neuroendocrine tumors [6, 7]. While measurement of 5-HIAA is useful in patients whose neuroendocrine tumors secrete serotonin, it has only limited utility in other contexts.

A need for new biomarkers in neuroendocrine tumors was highlighted as a key unmet need at consensus meetings sponsored by the National Cancer Institute in 2007 and 2011 and at a recent multinational expert consensus panel in 2015 [8-10]. MiRNAs are short (approximately 22nt’s) RNA sequences that have been shown to broadly regulate gene expression at a post-transcriptional level, by binding to the 3’ region of target RNAs, resulting in mRNA degradation and inhibition of translation [11]. MiRNAs are relatively stable in human tumor samples and can also be readily measured in blood specimens [12]. Their stability allows for the measurement of miRNAs across large numbers of specimens.

MiRNAs expression patterns furthermore appear to be specific to tumor type. In an initial study evaluating 217 miRNAs in 334 diverse human cancer samples, miRNA expression profiles were highly correlated with specific tumor lineages [13]. Measurement of miRNA expression patterns in blood specimens has also been shown to correlate with clinical outcomes. Studies in lung, NSCLC and ovarian cancer have identified specific serum miRNA signatures that correlate with survival [14-17]. The utility of miRNAs as biomarkers in patients with neuroendocrine tumors, however, remains uncertain [18]. Preliminary studies have suggested that miRNA patterns in neuroendocrine tumor tissue may correlate with tumor grade and stage [19, 20]. A more recent study revealed that, in patients with small intestine neuroendocrine tumors, tissue miRNAs can be detected in serum samples, and that serum levels may correlate with tumor stage and with treatment status [21].

To further define characteristic miRNAs in the serum of patients with neuroendocrine tumors, we undertook a multi-stage study; focusing on neuroendocrine tumors of the small intestine (SINET). We first developed a panel of candidate miRNAs detectable in patient plasma, based on evaluation of SINET samples and matched plasma samples. We refined our panel in an independent cohort of 40 cases and 40 controls and then validated this panel in a second, large cohort of 120 patients and 120 matched independent controls. Finally, we assessed whether the identified miRNAs in our panel were associated with clinical outcomes.

RESULTS

Identification of candidate miRNAs

To develop an initial candidate panel of characteristic miRNAs detectable in the plasma of patients with small intestine neuroendocrine tumors, we used a strategy in which we identified miRNAs similarly expressed in both tumor tissue and corresponding plasma samples. To perform this analysis, we identified 19 patients with small intestine neuroendocrine tumors, for whom both tumor tissue and paired plasma samples were available. Our screening cohort comprised 14 metastatic and 5 primary SINET cases (see Supplementary Table 1A).

We focused initially on analysis of the plasma samples. We profiled the 19 plasma samples from SINET patients, together with 19 age and gender-matched normal controls using the 742-miRNA panel (Exiqon, Inc). The panel contains many prominent miRNAs; however it is not exhaustive in terms of the total number of detectable microRNAs that have been discovered. We identified 64 miRNAs common to all plasma samples, with an average of 194 miRNAs detectable per sample. Expression levels were normalized to the mean expression level in the combined cohort (n=38). A supervised analysis of the ranked top 50 differentially expressed miRNAs demonstrated potential for clustering between cases and controls, suggesting the presence of characteristic miRNAs either over or under-expressed in SINET patients (Figure 1).

To further refine our list of candidate miRNAs, we next performed miRNA sequencing on the 19 paired tissue specimens. 519 miRNAs were detected in all tissue samples, with an average of 708 miRNAs detectable per sample. We normalized expression levels to the overall mean miRNA expression level in the 19 samples. We then correlated the miRNA expression data from the tumor samples with the miRNA profiles identified in the patient blood samples, with the goal of identifying characteristic miRNAs differentially expressed in a parallel fashion in both tissue and blood.

In identifying these candidate miRNAs, we limited our analysis to 184 miRNAs that were detected in at least 10/19 plasma and matching tumor tissue samples. We then binned normalized expression levels into quartiles for both data sets. Of the 184 miRNAs evaluated, 106 were concordant based on the binned score comparison as shown in Figure 2. In total 31 candidate miRNAs were identified (see Supplementary Table 2).
Figure 1: Heatmap and unsupervised hierarchical clustering. The heat map shows the result of the two-way hierarchical clustering of miRNAs and samples. Each row represents a single 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 single miRNA across all samples: red color represents an expression level above mean, green color represents expression lower than the mean. The clustering is performed on all samples, and on the top 50 miRNAs with highest standard deviation. Normalized (dCt) values have been used for the analysis (n=38).

Figure 2: Venn diagram comparison of binned scores from normalized generated from Exiqon profiling of blood and miRNA sequencing of matched tumor tissue from metastatic SINET patients.
Identification of an miRNA panel of characteristic miRNAs detectable in plasma of patients with metastatic small intestine NETs

To develop the final miRNA panel, we evaluated expression levels of the 31 previously identified candidate miRNAs in plasma samples from an independent group of 40 cases of metastatic SINETs and 40 normal healthy controls (see Supplementary Table 1B). Among the 31 candidate miRNAs, we identified 4 miRNAs; hsa-miR-22-3p, hsa-miR-21-5p, hsa-miR-29b-3p and hsa-miR-150-5p that were differently expressed between the patient and control groups at significance level of < 0.05. Levels of miR-21-5p and miR-22-3p were higher, and miR-29b-3p and miR-150-5p were lower in the plasma of the 40 metastatic small bowel NET patients compared to the 40 matched healthy controls (Figure 3A).

Validation of the 4-miRNA panel in cases and controls

We then validated miR-21-5p, miR-22-3p, miR-29b-3p and miR-150-5p in the plasma from a second, large independent cohort of 120 metastatic SINET patients and 120 matched normal healthy controls (see Supplementary Table 1C). As part of the experimental design we included 2 general tissue markers (miR-103-3p and miR-191-5p), 2 hemolytic markers (miR-23a-3p and miR-451a) and 2 internal normalization markers (miR-29c-3p and miR-425-5p). 9 SINET patient and 10 control samples were excluded from the analysis based on evidence of hemolysis. We confirmed that expression of 3 of the 4 miRNAs in our panel were significantly different between the patient and control groups. As in the previous cohort, we observed up-regulation of miR-21-5p and miR-22-3p and down-regulation of miR-150-5p in patients with metastatic SINETs (Figure 3B). Both miR-21 and miR-22 reside on chromosome 17 and are positively correlated with each other (r: 0.41, p<0.0001, Supplementary Table 3). miR-150 was strongly negatively correlated with miR-22 (r: -0.47, p<0.0001, Supplementary Table 3). The fourth miRNA, miR-29b-3p was downregulated compared to normal controls, though the difference did not reach statistical significance.

Association of the 4-miRNA panel with NET clinical features

To explore the potential utility of our 4 miRNA panel as a clinical tool, we first assessed the association of expression of the 4 miRNAs with standard clinical parameters in the 111 SINETs patients (Table 1).
Expression levels were dichotomized into low and high using the overall patient median as the cutoff. The median expression level was utilized as the cut-off to reduce possible data-driven bias. We found no differences in expression based on age at blood draw, patient gender and systemic treatment prior to blood draw (treatment naive or otherwise). Minor differences in expression levels were observed based on ethnicity (white or otherwise) and tumor differentiation; however results were not conclusive given the small size of the subgroups within the cohort. No significant difference was observed between the miRNA expression levels and whether patients were receiving treatment with a somatostatin analog.

### Association of miRNA expression and prognosis

To further assess the potential clinical relevance of the miRNAs identified in our study, we evaluated whether the level of miRNA expression in plasma was associated with overall survival (OS) in the validation cohort (n = 97 SINET patients, where follow-up was available). 14 SINET patients had no follow-up post-blood draw and were therefore excluded from the subsequent analysis. We assessed survival associations from time of blood draw to time of death or censoring, adjusting for age at blood draw, patient gender, ethnicity (white or otherwise), tumor differentiation (well or otherwise), and any met treatment prior to blood draw (treatment naive or otherwise). The median follow-up time was 4.3 years.

Using Kaplan-Meier and log rank methods, we found that low plasma expression of miR-21-5p (P=0.0103) and miR-22-3p (P=0.0155) and high expression of miR-150-5p (P=0.0368) were significantly associated with prolonged OS (Figure 4). We found that elevated plasma expression of miR-29b-3p was associated with a trend towards improved survival, though this did not reach statistical significance (P=0.1166). Multivariate analysis, as shown in Table 2, confirmed that high expression of plasma miR-21-5p and miR-22-3p was independently associated with shorter survival (miR-21-5p - HR: 0.54; 95% CI, 0.31-0.96; P=0.0345; miR-22-3p - HR: 0.56; 95% CI, 0.32-0.96; P=0.0362) whereas high expression of plasma miR-150-5p was associated with longer survival (HR: 2.24; 95% CI; 1.18-4.25; P=0.0136). Additionally we generated a new high/low risk index to combine miR-21-5p, miR-22-3p and miR-150-5p, considering high concentration for miR-21-5p and miR-22-3p and low concentration for miR-150-5p as high risk to further analyze their prognostic value as a 3-miR signature. The new index strengthened both the univariate (P=0.0025) and multivariate (P=0.0084) analysis. The combination of the 3 miRNA’s was associated with a
larger HR associated with shorter survival (HR: 0.47; 95% CI, 0.27-0.82) as shown in Figure 5. Elevated plasma CgA levels have been associated with poor overall prognosis in patients with advanced disease (Ter-Minassian, et al 2014). We evaluated the prognostic value in a model that combined CgA with the 3 miRNA’s signature, and observed an even stronger HR of 0.41 (95% CI, 0.2-0.85), with a significant association with shorter overall survival (P=0.0171). CgA concentration was dichotomized using 2x the upper limit of the range, (0.6-39 ug/L) as a cutoff for elevated expression.

We additionally assessed the 3 microRNA’s potential diagnostic utility. The highest AUC value from ROC estimates was calculated as 0.7025, when miR-21-5p, miR-22-3p and miR-150-5p were combined, with an ROC of 0.5 indicating no ability to discriminate between cases and controls, and an ROC of 1 indicating perfect discrimination between cases and controls.

**DISCUSSION**

In this study, we used a multi-step process to identify and validate characteristic miRNAs either over or under-expressed in the plasma of patients with metastatic small intestine NETs. Following the identification of 31 candidate miRNAs using paired plasma and tumor specimens, we identified 4 characteristically over or under-expressed miRNAs by comparing miRNA profiles in plasma from cases and normal controls. We then validated this panel in a separate, independent cohort of 111 cases and 110 controls. We have shown that elevated levels of miR-21-5p and miR-22-3p and low levels of miR-150-5p are found in metastatic small bowel neuroendocrine disease. We further found that expression of these miRNAs had prognostic significance: high plasma expression of miR-21-5p and miR-22-3p; and low expression of miR-150-5p were associated with poor overall survival.

![Figure 4: Kaplan-Meier survival product limit estimates for 97 SINET patients with metastatic disease in the validation cohort for plasma miR-21-5p, miR-22-3p, miR-29b-3p and miR-150-5p levels. The p-value was calculated using the log-rank test between patients with high and low expression.](image-url)
Table 2: Univariate and multivariate analyses of plasma miR-21-5p, miR-22-3p, miR-29b-3p and miR-150-5p association with overall survival in metastatic SINET patients

| MiRNA    | Expression | # Death/ at risk | Univariate analysis | Multivariate analysis | # Death/ at risk | Multivariate analysis |
|----------|------------|------------------|---------------------|-----------------------|------------------|-----------------------|
|          |            |                  | HR      | 95% CI | P-value | HR      | 95% CI | P-value |
| miR-21-5p| High       | 36/49            | ref     | ref    | ref     | 21/31  | ref    | ref     |
|          | Low        | 21/48            | 0.5     | 0.29-0.86 | 0.0118  | 0.54   | 0.31-0.96 | 0.0345 |
| miR-22-3p| High       | 33/49            | ref     | ref    | ref     | 21/35  | ref    | ref     |
|          | Low        | 24/48            | 0.53    | 0.31-0.89 | 0.0173  | 0.56   | 0.32-0.96 | 0.0362 |
| miR-150-5p| High      | 27/48            | ref     | ref    | ref     | 17/36  | ref    | ref     |
|          | Low        | 30/49            | 1.75    | 1.03-2.97 | 0.0392  | 2.24   | 1.18-4.25 | 0.0136 |
| miR-29b-3p| High      | 31/49            | ref     | ref    | ref     | 18/36  | ref    | ref     |
|          | Low        | 26/48            | 0.66    | 0.39-1.11 | 0.1192  | 0.63   | 0.36-1.08 | 0.0951 |
| 3-miR*   | High       | 35/50            | ref     | ref    | ref     | 21/33  | ref    | ref     |
|          | Low        | 22/47            | 0.44    | 0.25-0.75 | 0.0025  | 0.47   | 0.27-0.82 | 0.0084 |

*Combination includes high miR-21-5p, high miR-22-3p and low miR150-5p but not miR29b-3p.

Figure 5: Kaplan-Meier survival product limit estimates for 97 SINET patients with metastatic disease in the validation cohort for plasma combining miR-21-5p, miR-22-3p levels and miR-150-5p levels. The p-value was calculated using the log-rank test between patients with high and low expression. The high risk population includes patients with at least 2 positive out of the 3 biomarkers.
The identification of novel biomarkers in neuroendocrine tumors has proved challenging. Chromogranin A, a protein secreted from secretory granules in neuroendocrine cells, has become a standard biomarker in patients with neuroendocrine tumors despite having only limited utility. While chromogranin A has, in some studies, been associated with clinical outcomes, chromogranin A assays are highly variable [22, 23]. (Jensen, et al. 2013). Furthermore, levels of circulating Chromogranin A can be elevated in the setting of proton pump inhibitor use and in a number of non-malignant conditions including chronic renal failure and hepatic disease [3, 24]. Measurement of the serotonin metabolite 5-HIAA in blood or urine also has relatively limited utility, as only a minority of neuroendocrine tumors are associated with serotonin secretion.

Recent interest has focused on the use of blood-based markers based on neuroendocrine tumor genomics. A potential challenge with this approach is the relative genomic stability of neuroendocrine tumors, which precludes biomarkers based on the identification of characteristic genetic mutations. Whole genome profiling of small intestine neuroendocrine tumors revealed recurrent mutations in the cyclin dependent kinase inhibitor CDKN1B in only 8% of cases [25]. Pancreatic neuroendocrine tumors are also characterized by recurrent mutations in a relatively limited number of genes, which include tumor suppressor gene MEN1, as well as ATRX and DAXX, genes implicated in chromatin remodeling [26]. Preliminary studies have suggested that a multinanalyte approach, using a 51-transcript mRNA panel, may be more sensitive and specific than currently available biomarkers [27-29] though this assay has yet to be validated in large, prospective studies.

The measurement of circulating miRNAs represents a potentially promising approach for the development of more sensitive and specific biomarkers in neuroendocrine tumors. Initial tissue-based studies have suggested that neuroendocrine tumors have characteristic miRNA expression profiles. An initial analysis of 8 primary and 6 metastatic neuroendocrine tumors, using a 95-miRNA panel, suggested down-regulation of miR-133a in metastatic lesions [20] (Ruebel et al. 2010). Our miR-Seq results corroborate previous SINET-related miR tissue-based studies, where miR-133a has frequently been downregulated in the metastatic tumors in comparison to the primary tumors [19, 21, 30]. Miller also reported miR-1 and miR-143-3p were also downregulated and found NUAK2 and FOSB target genes were functionally crucial in the progression of SINETs. We too saw downregulation of miR-1 and miR-143-3p in our study in the metastatic tumors compared to the primary SINET tumors (data not shown).

Studies of circulating miRNAs in plasma however remain limited. Our observations that elevated levels of miR-21-5p and miR-22-3p and low levels of miR-150-5p are found in metastatic small bowel neuroendocrine disease but also have prognostic value support the further investigation of circulating miRNAs in neuroendocrine tumor disease. The observation that miR-21 is highly expressed in the plasma of patients with metastatic neuroendocrine tumors is consistent with its known function and with findings in other malignancies [2]. Expression of miR-21 has been shown to be associated with proliferation, apoptosis and migration [31] and involved in invasion, intravasation and metastasis [32, 33]. miR-21 is upregulated in glioblastoma [34], breast cancer [35, 36], lung cancer [37], gastric cancer [38], colorectal cancer [39], hepatocellular carcinoma [40], pancreatic cancer [41] as well as other tumor types [2]. Our finding that high expression of miR-21 is associated with shorter overall survival in metastatic small bowel NET patients, is also consistent with recent reports in metastatic breast [36], gastric, and cervical cancer [42, 43].

From the approximate 100 target genes that have been validated experimentally to be regulated by miR-21-5p (Supplementary Table 4), a number of key genes are known to be important in SINET tumorigenesis, including PDCD4, PTEN, AKT2, BCL-2 and SERPINB5. PTEN has an important role in NET biology, regulating the activity of MTOR through the AKT pathway, where its downregulated expression has been associated with shorter disease-free and overall survival [44, 45]. Suppressed PTEN has further been linked to miR-21 overexpression in pulmonary and pancreatic NETs [46-48]. Furthermore Chan et al reviewed the importance of the MTOR pathway in NET pathogenesis and reported on the testing of rapamycin and everolimus in Phase II trials [49]. PDCD4, located at 11q13 close to MEN1, has been found to be absent in NETs and has been shown to act as a tumor suppressor in neuroendocrine cells [50]. BCL-2, regulated by miR-21, was found to be predominantly negatively expressed in SINET tissues by IHC. Methylation of SERPINB5, another target gene of miR-21, is associated with reduced mRNA expression levels in relation to the loss of Chr18 in SINETs [51, 52]. We further found elevated expression of circulating miR-22 was associated both with the presence of metastatic small intestine NET and with shorter overall survival in this setting. The role of miR-22 in cancer development remains uncertain [53]. miR-22 is highly conserved across species, and has been shown to regulate phosphatase and tensin homologue tumor suppressor gene, PTEN [54]. MiR-22 was originally described as having a tumor suppressor function in several human malignancies, including breast [55], liver [56], colon [57] and ovarian cancer [58]. Other studies, however, have suggested it may also have an oncogenic function. Circulating miR-22 was found to be upregulated in NSCLC patients where it was also associated with disease progression [59]. miR-22 has also been shown to be upregulated and predictive in meningiomas and in prostate cancer [60, 61]. Our observations suggest that,
at least in neuroendocrine tumors, miR-22 may play
an oncogenic role. As with miR-21, high expression of
miR-22 was associated with shorter overall survival in
metastatic small bowel NET patients.

In contrast to our findings with miR-21 and miR-
22, we found that low (rather than high) levels of miR-
150 were characteristic of metastatic SINET and were
associated with shorter overall survival in this setting.
These findings are consistent with studies in other
malignancies, where miR-150 has predominantly been
described as having a tumor suppressor function. In
patients with hepatitis B virus, low expression of miR-
150 has been associated with a higher risk of HCC
development, as well as in shorter survival in patients
who develop HCC [62, 63]. In AML, lower levels of miR-150
are associated with disease recurrence [64].

While ROC curves suggested that the 3-miRNA
signature was likely to be of only limited value in
distinguishing neuroendocrine tumor cases from normal
controls, we did find that a model incorporating all three
miRNAs had strong prognostic significance. The model
was even stronger when CgA was taken into account,
suggesting that these miRNA signatures have the potential
to be used clinically as prognostic indicators. Given the
biological role of miRNAs as regulatory molecules, future
studies examining whether these signatures may also have
predictive value with regard to response to specific
systemic treatments are also warranted.

There are some limitations to our study. Firstly we
validated our panel in a large cohort of cases and controls
from a single institution; unencumbered by variation in
sample handling and processing. Further validation in
multi-institutional cohorts with standardized protocols
and inclusion of blood draws over the time course of a
therapeutic treatment regimen or in the setting of disease
progression would be warranted. Secondly there is no
single circulating miRNA/small RNA that on its own can
be utilized as the “housekeeping” reference across all
tumor types. Rather selection of an appropriate normalizer
needs to be undertaken for any given studied population
[65]. We tested a panel of 5 miRNAs as candidate
references, however were dissatisfied with the outcome
individually or in combination. We therefore elected
to use a global mean of all miRNAs being measured.
NormFinder confirmed that the global mean was the most
stable candidate reference within our cohorts. Thirdly
there is a lack of a real “normal” comparator tissue
applicable if Exiqon has performed sample prep). For
the RNA extraction of biofluids, 3 RNA spike-ins (UniSp2, UniSp4
and UniSp5) pre-mixed, each at different concentration
in 100 fold increments were added. This set of spike-
ins was intended as an RNA isolation control (only
applicable if Exiqon has performed sample prep). For
the reverse transcription step, one spike-in (UniSp6)
was added. This control was used to confirm that the
reverse transcription and amplification occurs with equal
efficiency in all samples. Results were reported as a Ct,
which is calculated, as the maximum second derivative
of the amplification curve. For generating ΔCt values, we
utilized a global mean Ct value of all miRNAs measured
in the original screen (742 miRNA), the follow-up target panel

MATERIALS AND METHODS

Patient and control samples

Neuroendocrine tumor tissue and plasma samples
were obtained from the Dana-Farber Cancer Institute
Neuroendocrine Tumor Biospecimen database, under an
IRB-approved protocol in which patients provide informed
consent for collection of tissue specimens obtained during
routine clinical care, as well as collection of additional
blood specimens for research use. Plasma was also
collected from consenting spousal/friend as healthy
controls. Briefly blood (up to 10ml) was collected into
tubes containing EDTA, and the separation procedure was
carried out within 3hr of venipuncture. The samples were
spun at 3000rpm for 15min resulting in approximately
3-6ml plasma. The plasma was aliquoted and stored
at -80°C for preservation. A portion of each tissue
sample was preserved in OCT and evaluated to confirm
histological features and tumor cellularity. Additional
portions of each tissue were snap frozen and banked for
subsequent molecular characterization.

miRNA profiling in plasma samples

miRNA profiling in plasma samples was performed
using the miRCURY LNA (Locked Nucleic Acid)
Universal miRNA PCR assays [Exiqon, Vedbaek,
Denmark], where all miRNAs are polyadenylated
and reverse transcribed into cDNA in a single step and
amplification is performed on a Roche Lightcycler 480.
A total of 742 miRNAs were assayed. For the RNA
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of the amplification curve. For generating ΔCt values, we
utilized a global mean Ct value of all miRNAs measured
in the original screen (742 miRNA), the follow-up target panel

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(31 miRNAs) and the validation set. NormFinder was utilized to identify the optimal normalization miRNA(s) among a set of candidate references [66]. No template controls (NTC’s) were included as negative controls. All Ct values within 5 Ct values of NTC’s were automatically omitted from all analysis. The $2^{-\Delta\Delta Ct}$ method was used to analyze the relative expression of miRNAs, and the student t-test to compare the expression between the patient and control groups for the original screen and the follow-up target panel analysis. To monitor hemolysis two miRNAs were used; one that is expressed in red blood cells (miRNA-451) and one that is relatively stable in serum and plasma and is not affected by hemolysis (miRNA-23a).

**miRNA sequencing in tumor samples**

Total RNA, enriched for miRNA extractions were performed on snap frozen tissue specimens. H&Es were reviewed to identify enriched regions of high tumor cellularity, although tumors were not microdissected to select only for tumor cells exclusively. Briefly snap frozen tumor was cut and processed using a QIAshredder [Qiagen #79654], and miRNA-enriched total RNA was extracted using MiRNeasy Mini Kit [Qiagen # 217004]. 1μg of RNA was used as input in the ScriptMiner™ Small RNA-Seq Library Preparation Kit [EpiCenter #SMMP101212]. Resulting cDNA libraries had a maximum peak of approximately 140bp. Libraries were QC’d with a Agilent High Sensitivity DNA Kit on a 2100 bioanalyzer system [Agilent #5067-4626]. qPCR library quantification was performed using KAPA SYBR® FAST qPCR Kits [KapaBiosystems, Wilmington, MA]. A final concentration of 10pM was loaded onto v3 Illumina flow cells. Libraries were sequenced as 50bp single end reads on a HiSeq 2200 system. Sequenced reads were aligned to the reference human genome (hg19) using the universal RNA-Sq aligner, STAR algorithm to produce BAM files. MiRNAs were quantified in the sample using the mirDeep2 algorithm, which produces count values for all the known miRNA.

**Statistical analysis for validation**

Non-parametric Wilcoxon rank sum test was performed to compare the patient and control groups. Descriptive statistics of patient clinical features were tabulated by the dichotomized expression level of 4-miRNA panel at median, and compared by Wilcoxon rank sum test, chi-square test, or fisher's exact test when applicable. The distribution of survival times by dichotomized 4-miRNA panel were assessed using Kaplan-Meier methods and log-rank tests, and then the association was evaluated by Cox proportional hazards regression. Multivariate models were adjusted for age and blood draw, gender, race (white or otherwise), tumor differentiation, and treatment naive or otherwise prior to blood draw. Proportional hazard assumptions were verified by testing for a nonzero slope of scaled Schoenfeld residuals on ranked failure times. In addition, Pearson's correlation was applied to determine the association between all pairs in the 4-miRNA panel. Analyses were conducted using SAS version 9.4 (SAS Institute, Cary, NC). Two-sided p-values <0.05 were considered statistically significant.

**MiRNA target database**

The miRTarBase (Release 6.0 September 2015) was utilized to assess the miRNA-21-5p - target interactions (MTIs). 99 genes were found to have been validated experimentally by reporter assay, western blot and microarray, which together are characterized as strong evidence of an interaction [67].

**Abbreviations**

- 5-HIAA: 5-Hydroxyindoleacetic acid; SINET: small intestine neuroendocrine tumor; EDTA: Ethylenediaminetetraacetic acid; OCT: optimal cutting temperature compound; LNA: locked nucleic acid; PCR: polymerase chain reaction; BAM: binary alignment/map; OS: overall survival; HR: hazard ratio; AUC: area under the curve; ROC: receiver operating characteristic; NSCLC: non-small cell lung carcinoma; AML: acute myeloid leukemia; HCC: hepatocellular carcinoma; CgA: chromogranin A.

**Author contributions**

MK is the senior author of this manuscript. AT, MK and MB supervised the project and developed concepts. AT, MK, and MB designed experiments. AT, LN, MK and MB interpreted data. AT, MK and MB gave conceptual advice and edited the manuscript. CWZ and AR performed extractions and small RNA library prep. WK performed unsupervised hierarchical clustering. SZ performed statistical analysis of the validation data set. MK and MB wrote the manuscript. LB and ES retrieved tissue and compiled clinical information and demographics.

**CONFLICTS OF INTEREST**

The authors declared that there are no conflicts of interest in this work.
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