## Research Article

# Uncontrolled Diabetes Mellitus Has No Major Influence on the Platelet Transcriptome

Thomas G. Nührenberg <sup>(b)</sup>,<sup>1,2</sup> Marco Cederqvist,<sup>2</sup> Federico Marini,<sup>3</sup> Christian Stratz,<sup>2</sup> Björn A. Grüning,<sup>4</sup> Dietmar Trenk,<sup>2</sup> Harald Binder,<sup>5</sup> Ralf Gilsbach,<sup>1</sup> Franz-Josef Neumann,<sup>2</sup> and Lutz Hein <sup>(b)</sup>,<sup>1,6</sup>

<sup>1</sup>Institut für Experimentelle und Klinische Pharmakologie und Toxikologie, Albert-Ludwigs-Universität Freiburg, D-79104 Freiburg, Germany

<sup>2</sup>Universitäts-Herzzentrum Freiburg, Bad Krozingen, Abteilung für Kardiologie und Angiologie II, D-79189 Bad Krozingen, Germany

<sup>3</sup>Institut für Medizinische Biometrie, Epidemiologie und Informatik, Universitätsmedizin der Johannes-Gutenberg-Universität Mainz, D-55101 Mainz, Germany

<sup>4</sup>Institut für Informatik, Albert-Ludwigs-Universität Freiburg, D-79110 Freiburg, Germany

<sup>5</sup>Institut für Medizinische Biometrie und Statistik (IMBI), Albert-Ludwigs-Universität Freiburg, D-79104 Freiburg, Germany <sup>6</sup>BIOSS Centre for Biological Signaling Studies, University of Freiburg, D-79104 Freiburg, Germany

Correspondence should be addressed to Thomas G. Nührenberg; thomas.nuehrenberg@universitaets-herzzentrum.de and Lutz Hein; lutz.hein@pharmakol.uni-freiburg.de

Received 27 May 2018; Revised 26 September 2018; Accepted 11 October 2018; Published 1 November 2018

Academic Editor: Graziano Pesole

Copyright © 2018 Thomas G. Nührenberg et al. This is an open access article distributed under the Creative Commons Attribution License, which permits unrestricted use, distribution, and reproduction in any medium, provided the original work is properly cited.

*Background.* Diabetes mellitus (DM) has been associated with increased platelet reactivity as well as increased levels of platelet RNAs in plasma. Here, we sought to evaluate whether the platelet transcriptome is altered in the presence of uncontrolled DM. *Methods.* Next-generation sequencing (NGS) was performed on platelet RNA for 5 patients with uncontrolled DM (HbA1c 9.0%) and 5 control patients (HbA1c 5.5%) with otherwise similar clinical characteristics. RNA was isolated from leucocyte-depleted platelet-rich plasma. Libraries of platelet RNAs were created separately for long RNAs after ribosomal depletion and for small RNAs from total RNA, followed by next-generation sequencing. *Results.* Platelets in both groups demonstrated RNA expression profiles characterized by absence of leukocyte-specific transcripts, high expression of well-known platelet transcripts, and in total 6,343 consistently detectable transcripts. Extensive statistical bioinformatic analysis yielded 12 genes with consistently differential expression at a lenient FDR < 0.1, thereof 8 protein-coding genes and 2 genes with known expression in platelets (*MACF1* and *ITGB3BP*). Three of the four differentially expressed noncoding genes were YRNAs (*RNY1, RNY3, and RNY4*) which were all downregulated in DM. 23 miRNAs were differentially expressed between the two groups. Of the 13 miRNAs with decreased expression in the diabetic group, 8 belonged to the DLK1–DIO3 gene region on chromosome 14q32.2. *Conclusions*. In this study, uncontrolled DM had a remote impact on different components of the platelet transcriptome. Increased expression of *MACF1*, together with supporting predicted mRNA-miRNA interactions as well as reduced expression of RNYs in platelets, may reflect subclinical platelet activation in uncontrolled DM.

## 1. Introduction

Patients with diabetes mellitus are at high risk for cardiovascular events such as myocardial infarction or stroke. Blood platelets play a pivotal role in hemostasis and thrombus generation and have been shown to be hyperreactive in patients with diabetes mellitus [1, 2]. Increased calpain activation has been identified as one mechanism leading to enhanced secretion of platelet  $\alpha$ -granule proteins [3] which in turn might promote atherosclerosis in diabetic patients. Aside from various cytokines stored in  $\alpha$ -granules as well as small molecules such as ATP, ADP, and serotonin in

dense granules, platelets contain abundant microRNAs or miRNAs [4, 5]. These short, noncoding RNAs are under intense clinical and experimental investigation due to their potential as biomarkers and as intercellular messengers [1]. Platelet miRNAs and mRNAs can be secreted within vesicles [2–4] and lead to changes in gene expression of endothelial cells.

With regard to diabetes mellitus, Zampetaki et al. reported an unexplained loss of several endothelial- and platelet-derived miRNAs in the plasma of diabetic patients [5]. On the other hand, more recent data from the same group clearly indicate that levels of plasma miRNAs largely correlate with platelet miRNAs and in part with platelet activation levels [6, 7]. Disease-specific differences in plasma miRNAs should therefore either be due to altered release from platelets or altered clearance from the plasmatic compartment. Assessing the former possibility, our group has analyzed platelet miRNAs in a cohort of 60 patients with diabetes mellitus [8]. Compared to nondiabetic patients, patients with wellcontrolled disease had no major changes in the platelet miRNA profile. However, potential differences mediated by the presence of diabetes mellitus might have been attenuated by the overall good glucose control in the diabetic patients.

We therefore sought to further investigate the RNA profile of platelets from patients with uncontrolled diabetes mellitus in comparison to matched control patients without diabetes mellitus. In the current study, we used next-generation sequencing of both long and short RNAs, followed by comprehensive bioinformatic analysis.

#### 2. Material and Methods

2.1. Subjects, Exclusion Criteria, and Blood Sampling. Blood was drawn from 10 patients that were electively hospitalized at the Universitäts-Herzzentrum Freiburg • Bad Krozingen and in stable clinical condition. Patients with unstable angina pectoris, myocardial infarction with or without ST-elevation, significant renal dysfunction, malignancy and pathological platelet count were excluded. Patients with uncontrolled diabetes were required to have an HbA1c of 8.5% or greater at the time of blood sampling. For the control group, patients with similar clinical characteristics but absence of diabetes were selected. For platelet isolation, blood was collected into 7.5 ml trisodium citrate (0.106 mol / l) monovettes (Sarstedt, Nümbrecht, Germany) after peripheral venous puncture with a safety Multifly-Set® (Sarstedt, Nümbrecht, Germany) and immediately processed.

All participants gave written informed consent to participate in the study which was approved by the Ethics Committee of the Albert-Ludwigs-Universität Freiburg. The investigation conformed to the principles outlined in the Declaration of Helsinki.

2.2. Platelet Function Tests. Light transmission aggregometry (Platelet Aggregation Profiler PAP-4 Bio/Data Corporation Horsham, PA, USA) using adenosine diphosphate (ADP  $5\mu$ M and  $20\mu$ M), Thrombin Receptor Activating Peptide (TRAP 25  $\mu$ M), arachidonic acid (500mg/l), and collagen (1mg/l

and 2.5 mg/l, collagen fibrils from equine tendo) as platelet activators was carried out on platelet-rich plasma (PRP) to analyse basic functional platelet properties and to evaluate the platelet reaction to acetylsalicylic acid. Furthermore, multiplate electric impedance aggregometry was performed in whole blood collected into 2.7 mL tubes containing rhirudin ( $45\mu$ g/mL, Sarstedt, Nümbrecht, Germany) using the Multiplate Analyzer<sup>®</sup> (Roche Diagnostics, Mannheim, Germany). TRAP, ASPI, ADP, and hsADP tests were performed according to manufacturers' instructions: recording the area under the curve of aggregation units over 6 minutes. Results are reported as aggregation units \* minute (AU \* min).

2.3. Purification of Platelets by Leukocyte Depletion and RNA Isolation. Platelet purification and leukocyte depletion were performed as previously described [8, 9]. In brief, platelet rich plasma (PRP) was obtained by centrifugation (Megafuge 1.0 RS; Heraeus, Hanau, Germany) of citrated blood samples at 2040 rpm (750 rcf) for 2 min without brake. Depletion of leukocytes was achieved by negative selection using CD45<sup>+</sup> magnetic beads (EasySep<sup>™</sup>; STEMCELL Technologies, Köln, Germany) according to the manufacturer's instructions. Total RNA including miRNA was extracted from the leukocyte depleted platelet pellets using the miRNeasy Mini Kit (Qiagen GmbH, Hilden, Germany) according to the manufacturer's recommendations. The eluted RNA was stored at -80°C.

2.4. Library Preparation and Next-Generation Sequencing. RNA yield was determined using an Agilent 2100 bioanalyzer with the eukaryote total RNA pico assay (Agilent, Waldbronn, Germany). In average  $26.8 \pm 7.4$  ng total RNA were subjected to depletion of ribosomal RNA using the RiboMinus™ Eukaryote System v2 (life technologies, Darmstadt, Germany) according to manufacturers' instructions and directly used for preparation of 10 libraries with the NEBNext® Ultra<sup>™</sup> Directional RNA Library Prep Kit for Illumina® (New England BioLabs Inc., Frankfurt am Main, Germany). Library size was assessed before sequencing using the Agilent 2100 bioanalyzer high sensitivity DNA analysis kit. In total, 6 small RNA libraries were successfully created from in average 84.4 ± 19.1 ng of total RNA with NEBNext® Small RNA Library Prep Set for Illumina® (New England BioLabs Inc., Frankfurt am Main, Germany). Barcoded libraries were sequenced on an Illumina HiSeq 2500.

2.5. Statistical Analysis and Bioinformatic Analysis. All sequencing data have been deposited in the European nucleotide archive (ENA; http://www.ebi.ac.uk/ena) under the accession number PRJEB18466. The datasets were independently analyzed by TN and FM at the respective locations. Both datasets were aligned to the hg38 version of the human genome, either by the Tophat2 (TN) [10] or by STAR aligner (FM) [11]. Data were analyzed for different gene expression with HTSeq [12] and DESeq2 [13] either without further data reduction or after removal of putative PCR duplicates by picard [https://github.com/broadinstitute/picard] or SAM-tools, respectively [14]. Genes were considered differentially expressed at an FDR of <0.1 if they were identified by

at least two of the three analysis conditions. This avoids artifacts from specific analysis pipelines, while at the same time the lenient level of 0.1 reflects the idea that potential differences should not be missed. Datasets for small RNA-Seq were analyzed similarly, but without duplicate removal. In addition, both datasets were analyzed in a similar fashion with more recent tools for detection of differential gene expression such as edgeR incorporating a quasilikelihood approach (edgeR-QLF) [15] and RNentropy [16] to further evaluate the robustness of results. To account for factors that might affect gene expression aside from Diabetes mellitus, we evaluated the effects of clinical characteristics on the principal components of gene expression, in such a way that important characteristics could subsequently be entered as covariates to the design of the experiment in the generalized linear model framework provided by DESeq2. If not indicated otherwise, clinical data are presented with median and interquartile range in brackets.

2.6. MiRNA-mRNA Interactions and Target Prediction. Differentially expressed miRNAs and mRNAs were analysed by Magia [2], a web tool for integrated analysis of mRNAs and miRNAs [17]. In detail, the prediction tools Targetscan [18], PicTar [19], and RNA22 [20] were used in complementary fashion.

#### 3. Results

*3.1. Baseline Characteristics.* The 5 diabetic and 5 nondiabetic patients were well matched with regard to clinical and laboratory parameters as well as medication (Table 1). Concerning HbA1c, diabetic patients showed a median of 9.0% [8.5-9.9%] compared to 5.5% [5.4-5.7%] in the control group as intended. C-reactive protein was also significantly increased in diabetic patients (0.8 [0.6-1.2] mg/dl versus 0.3 [0.1-0.3] mg/dl), with values slightly above normal reference values (Table 1).

3.2. Multiplate Electric Impedance and Light Transmission Aggregometry. Multiplate electric impedance aggregometry with routine stimulation protocols showed no major differences in aggregation characteristics between the two groups. Light transmission aggregometry revealed decreased aggregation in diabetic patients upon low dose collagen stimulation and a trend to less aggregation upon stimulation with TRAP (Table 2).

*3.3. Platelet Purification and Leukocyte Depletion.* Analysis of platelet and leukocyte marker transcripts confirmed the high degree of purity of the platelet preparations. The chemokine *CCL5* was highly expressed in platelets (Figure 1(a), left panels), whereas leukocyte markers, including CD45 (*PTPRC*), were not detectable (Figure 1(a), right panels).

3.4. Detection of Platelet Transcripts by Long RNA Sequencing and Influence of Uncontrolled Diabetes Mellitus. Mapping of sequenced ribosome-depleted RNA libraries with Tophat2 resulted in a median of 35.8 [32.7-39.0] million uniquely mapped reads. 5898 unique transcripts were consistently expressed in all samples. Of these, 5239 were protein-coding genes, 365 pseudogenes and 104 lincRNAs (Figure 1(b)). Regarding differential expression in diabetes mellitus, only 12 transcripts were detected as differentially expressed. Of these genes, 8 were protein-coding genes whereas 4 transcripts were noncoding RNAs (Table 3, Figure 2(a)). Adjustment for several clinical variables according to principal component analyses did not yield consistent effects of clinical characteristics on gene expression. Therefore, no adjustment was performed. Additional analysis with RNentropy did not yield any differentially expressed gene while comparison of DESeq2 with edgeR-QLF resulted in a large overlap of differentially expressed genes (Table 3, Figure 2(a), Supplementary table 1).

3.5. Differences in Short Platelet RNA Sequencing in Diabetic versus Non-Diabetic Patients. In the short RNA sequencing data, 445 RNAs were consistently expressed in all samples. Of these, 380 belonged to miRNAs, 62 to fragmented YRNAs, and RNYs and 3 to vault RNAs (Figure 1(c)). 24 short RNAs thereof 23 miRNAs were differentially expressed between diabetic patients and controls. Of the 13 miRNAs with decreased expression in the diabetic group, 8 belonged to the DLK1–DIO3 gene region on chromosome 14q32.2 (Table 3). Additional analysis with edgeR-QLF identified differentially expressed miRNAs, with complete overlap in miRNAs with decreased expression in the diabetic group and DLK1–DIO3 gene region (Table 3, Supplementary table 2).

3.6. Differentially Expressed miRNAs and mRNAs form a Putative Regulatory Network. miRNA target analysis by Magia [2] revealed a network including 4 of the 8 differentially expressed protein-coding genes. In particular, 4 of the 6 protein-coding genes with increased expression in diabetes mellitus belonged to the network. All 12 predicted positive or negative miRNA correlations were in line with the observed increased expression of the 4 protein-coding genes (Figure 2(b)).

#### 4. Discussion

As main finding, no major differences are found in small and long RNAs in platelets of diabetic patients with poor glycemic control, compared to those from nondiabetic controls with similar clinical characteristics.

Few studies have measured the expression of platelet miRNA in diabetic patients and non-diabetic controls [8, 21]. Elgheznawy et al. reported a reduced expression of 4 miRNAs in platelets from eight diabetic patients compared to age-matched healthy controls [21]. In this study, no leukocyte-depletion of platelets was performed. This is relevant since miRNA expression was normalized to 18S ribosomal RNA which is highly expressed in nucleated cells such as leukocytes. In this context, we and others have shown that diabetic patients exhibit higher leukocyte counts [8, 22] than nondiabetic controls. Therefore, it cannot be excluded that higher leukocyte contamination explains the described differences. Another group has recently reported

	Diabetes	no Diabetes	p value	
Male sex	3 (60%)	3 (60%)	1	
Age (years)	70 [63-72]	63 [59-70]	0.492	
BMI $(kg/m^2)$	33.4 [32.0-35.3]	31.8 [30.5-38.8]	0.841	
Hypertension	5 (100%)	4 (80%)	1	
Active smoking	1 (20%)	1 (20%)	1	
Coronary artery disease	4 (80%)	2 (40%)	0.524	
Remote coronary intervention	3 (60%)	2 (40%)	1	
Reduced left ventricular function	2 (40%)	2 (40%)	1	
Medication				
Insulin	3 (60%)	0 (0%)	0.167	
Oral antidiabetics	3 (60%)	0 (0%)	0.167	
ASA	3 (60%)	3 (60%)	1	
Oral anticoagulation	3 (60%)	2 (40%)	1	
Statins	3 (60%)	2 (40%)	1	
Laboratory parameters				
Platelets $(10^3/\mu l)$	236 [183-238]	212 [207-225]	1	
Leukocytes $(10^3/\mu l)$	9.0 [6.3-9.2]	8.6 [6.2-9.0]	0.548	
Hemoglobin (g/dl)	13.4 [12.6-13.8]	14.7 [14.3-14.8]	0.143	
Total cholesterol (mg/dl)	163 [160-168]	178 [168-219]	0.111	
LDL cholesterol (mg/dl)	104 [65-106]	126 [114-128]	0.056	
HDL cholesterol (mg/dl)	38 [29-41]	49 [42-87]	0.111	
Triglycerides (mg/dl)	203 [100-280]	150 [118-151]	0.690	
Serum glucose (mg/dl)	255 [202-256]	97 [87-107]	0.151	
HbA1c (%)	9.0 [8.5-9.9]	5.5 [5.4-5.7]	0.008	
C-reactive protein (mg/dl)	0.8 [0.6-1.2]	0.3 [0.1-0.3]	0.048	
Creatinine (mg/dl)	1.0 [0.9-1.5]	0.9 [0.8-1.0]	0.444	

TABLE 1: Baseline characteristics of the patients. BMI: body mass index and ASA: acetylsalicylic acid.

TABLE 2: Multiplate electric impedance and light transmission aggregometry. TRAP: thrombin receptor-activating peptide, ADP: adenosinediphosphate, HS: high-sensitivity, and AU: area under the curve.

	Diabetes	no Diabetes	p (Mann-Whitney-U-test)
Multiplate electric impedance aggregometry			
TRAPtest (AU*min)	748 [721-1241]	941 [924-1019]	0.690
ADPtest (AU*min)	707 [520-741]	696 [692-718]	0.841
ADPtest HS (AU*min)	607 [400-610]	509 [461-568]	1
ASPItest (AU*min)	342 [188-346]	413 [255-599]	0.518
Light transmission aggregometry			
TRAP 25µM (% aggregation)	85 [84-86]	100 [100-100]	0.087
ADP $5\mu$ M (% aggregation)	47 [34-49]	66 [44-72]	0.310
ADP 20µM (% aggregation)	82 [67-84]	88 [74-94]	0.548
Arachidonic acid 500mg/l (% aggregation)	7 [7-10]	9 [6-96]	0.802
Collagen 1mg/l (% aggregation)	10 [10-11]	18 [14-74]	0.016
Collagen 2.5mg/l (% aggregation)	28 [23-43]	60 [39-100]	0.103

moderate reductions of 4 platelet miRNAs from diabetic patients [23]. In this study, qPCR assays were normalized to a small nucleolar RNA that could also easily be influenced by very low amounts of contaminating leukocytes [23].

We have previously reported the absence of major changes in the platelet miRNA profile in a well-matched cohort of sixty patients when leukocyte-depletion and array-technology was applied [8]. The current data, again with proof of leukocyte depletion, extend these findings to the entire transcriptome in individuals with poor glycemic control.

The few specific differences found in mRNA and miRNA expression yielded a consistent regulatory network, which is highly suggestive of persistent miRNA-mRNA target

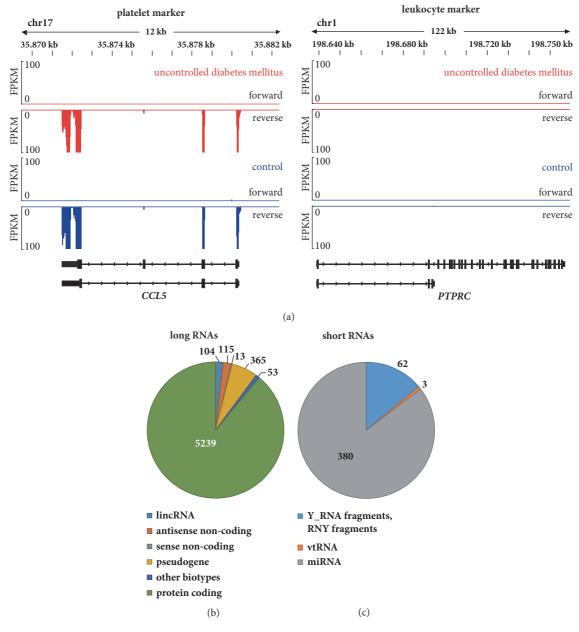


FIGURE 1: **Purity and gene biotypes of platelet RNAs**. (a) Genome browser view for transcripts of the platelet-enriched gene *CCL5* (left panel) and the leukocyte-specific gene *PTPRC* (CD45, right panel). FPKM denotes fragments per kilobase of transcript per million mapped reads. (b) Gene biotypes for mapped long RNAs. (c) Gene biotypes for mapped short RNAs.

interactions in platelets. The observed increased expression of *MACF1* might be relevant in the context of platelet activation. *MACF1* is involved in the cross-linking of microtubules and actin filaments [24], structural components important for maintenance of the discoid platelet shape [25, 26], and granule secretion [27]. Furthermore, reduced expression of *ITGB3BP*, also known as  $\beta$ 3-endonexin, might prevent platelet aggregation since *ITGB3BP* increases the affinity of integrin  $\alpha_{\text{IIb}}\beta_3$  for fibrinogen binding [28]. Yet, the relevance of these observations for platelet function is currently unknown. It is well possible that the observed changes are only a reminiscence of changes that are already present in the

megakaryocyte. As platelets lack genomic DNA, two of the differentially expressed genes, *UHRF1BP1* and *HIST1H2BF*, may indeed reflect transcriptome differences between diabetic and non-diabetic megakaryocytes. *HIST1H2BF* encodes for a histone H2B variant, which represents an essential chromatin component. UHRF1BP1 interacts with DNA methyl-transferases and may thus modulate DNA methylation, gene expression, and cellular differentiation.

A recent report has shown that translation in platelet is confined to the first hours of the platelet life span [29]. One could therefore speculate that individuals with higher platelet turnover dispose of platelets that are more

TABLE 3: Differential short RNA expression in platelets from patients with uncontrolled diabetes mellitus. + marks miRNAs that belong to the DLK1–DIO3 gene region on chromosome 14q32.2. FDR: Bonferroni-Hochberg-adjusted False-Discovery-Rate. miRNAs in bold were also differentially expressed by analysis with edgeR-QLF.

	gene biotype	<b>log</b> <sub>2</sub> <b>fold change</b> (Diabetes vs. Control)	FDR
hsa-miR-106b-3p	miRNA	0.91	0.07
hsa-miR-143-5p	miRNA	1.73	0.06
hsa-miR-145-5p	miRNA	1.90	0.06
hsa-miR-152-3p	miRNA	0.75	0.07
hsa-miR-185-5p	miRNA	0.84	0.07
hsa-miR-326	miRNA	1.06	0.08
hsa-miR-425-3p	miRNA	0.73	0.09
hsa-miR-652-3p	miRNA	0.85	0.01
hsa-miR-671-5p	miRNA	1.27	0.07
hsa-miR-93-5p	miRNA	0.77	0.06
hsa-let-7f-5p	miRNA	-0.79	0.09
hsa-miR-128-3p	miRNA	-1.65	0.00
hsa-miR-26a-5p	miRNA	-1.18	0.07
hsa-miR-340-3p	miRNA	-1.45	0.03
+ hsa-miR-493-5p	miRNA	-1.25	0.07
+ hsa-miR-411-3p	miRNA	-1.31	0.06
+ hsa-miR-543	miRNA	-1.45	0.06
+ hsa-miR-495-3p	miRNA	-1.06	0.08
+ hsa-miR-654-3p	miRNA	-1.08	0.08
+ hsa-miR-539-3p	miRNA	-2.18	0.02
+ hsa-miR-485-3p	miRNA	-1.17	0.07
+ hsa-miR-369-5p	miRNA	-1.55	0.00
Y_RNA.518-20-3p	misc RNA	-1.77	0.07

adaptive to pathologic activation. Yet, the presence of higher platelet turnover in patients with diabetes mellitus remains to be shown. Regarding platelet function testing, no difference between diabetic and nondiabetic patients could be observed. Given the correlation of RNY fragments in plasma originating from platelets and platelet reactivity, RNY in plasma might be a more sensitive marker of platelet activation than classic platelet function tests [6]. This notion is underlined by our finding of reduced RNY expression in platelets from patients with uncontrolled diabetes mellitus, suggestive of increased secretion of RNY into plasma. Noteworthy, a substantial proportion of downregulated miR-NAs belonged to the DLK1-DIO3 gene region on chromosome 14q32.2. Edelstein et al. showed that black Americans with increased PAR4-mediated platelet reactivity have also decreased expression of miRNAs within the DLK1-DIO3 gene region [30]. Similarly, miRNAs of this gene region were epigenetically repressed in islet cells from diabetic patients [31]. It therefore appears that downregulation of these miRNAs may be a general feature of type II diabetes but may indirectly influence platelet reactivity.

4.1. Limitations. One limitation of the present study is its small sample size. However, use of next-generation sequencing allowing precise determination of genome-wide presence of short and long transcripts precludes selection or

normalization bias that is inherent in other array- or qPCRbased techniques. Certainly, minor differences could have been substantiated by enlarging the sample size. Yet, we already chose a lenient level FDR, to counter potential lack of power due to sample size. In addition, we employed different tools for detection of differential gene expression which may help to reduce the number of false positives. The identified set of differentially expressed genes in diabetic platelets will be an important basis for validation in a larger cohort.

Recently, the presence of circular RNAs in platelets was reported pointing to a degraded transcriptome [32]. An analysis using PTESFinder [33] with regard to differences in diabetic patients was not meaningful due to the sequencing length of 50 bp in our study (as compared to at least 75 bp recommended sequencing length for computational identification of circRNAs). In addition, we did not detect an increased platelet reactivity in diabetic patients using the Multiplate impedance aggregometry which is recommended by current guidelines for platelet function testing [34]. Flow cytometric analysis of monocyte-platelet aggregates assessing low-level platelet activation may have established subclinical platelet activation in diabetic patients.

As another limitation, we did not include a control group of patients with well-controlled diabetes. Hence, we cannot differentiate whether the observed changes depend on glycemic control or are driven by other factors present

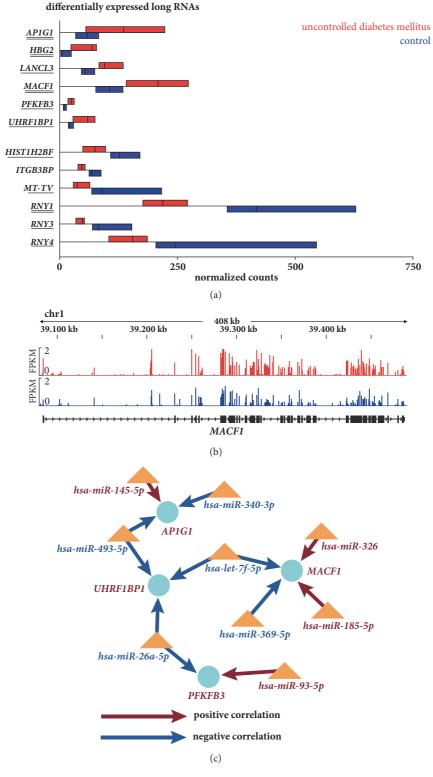


FIGURE 2: Differential long RNA expression and mRNA-miRNA interaction network. (a) Visualization of differentially expressed long RNAs by floating bars depicting the range of normalized counts from minimal to maximal values within in the respective groups. Black lines within the boxes represent the median. Red boxes show counts for transcripts of patients with uncontrolled diabetes mellitus, blue boxes show values of control patients. Underlined genes were also differentially expressed by analysis with edgeR-QLF, either after duplicate removal with picard (single line) or also with SAMtools (double line). (b) Genome browser view for transcripts of the *MACF1* gene. Transcripts of patients with uncontrolled diabetes mellitus are shown in red, transcripts of control patients in blue. FPKM denotes fragments per kilobase of transcript per million mapped reads. (c) Visualization of predicted mRNA-miRNA interactions obtained by analysis with Magia [2]. mRNAs are depicted by blue circles, miRNAs by orange triangles. Upregulation of mRNAs or miRNAs in uncontrolled diabetes mellitus is shown in red, downregulation in blue. The predicted direction of interaction is represented by the color of the arrows.

in diabetic patients. Yet, we did not see pronounced changes in miRNA levels of patients with controlled diabetes as compared to patients without diabetes previously [8].

4.2. Conclusions. In summary, uncontrolled diabetes is associated with minor differences in short and long platelet RNAs. Increased expression of *MACF1*, together with supporting predicted mRNA-miRNA interactions as well as reduced expression of RNYs in platelets, may reflect subclinical platelet activation in uncontrolled diabetes mellitus.

#### Abbreviations

MiRNA: MicroRNA PRP: Platelet-rich plasma RNA-Seq: RNA-Sequencing.

#### **Data Availability**

All sequencing data have been deposited in the European nucleotide archive (ENA; http://www.ebi.ac.uk/ena) and is available under the accession number PRJEB18466.

#### Disclosure

Current address of Christian Stratz is Novartis AG, Basel, Switzerland.

## **Conflicts of Interest**

No conflicts of interest are declared for this article.

## **Authors' Contributions**

Thomas Nührenberg, Marco Cederqvist, and Christian Stratz performed experiments; Thomas Nührenberg, Federico Marini, Björn Grüning, and Harald Binder analysed data; Harald Binder interpreted results of experiments; Thomas Nührenberg and Federico Marini prepared figures; Thomas Nührenberg, Marco Cederqvist, Federico Marini, Harald Binder, Ralf Gilsbach, Dietmar Trenk, Franz-Josef Neumann, and Lutz Hein edited and revised manuscript; Thomas Nührenberg, Franz-Josef Neumann, and Lutz Hein had the concept of the research and designed it; Thomas Nührenberg and Christian Stratz drafted the manuscript. All authors have read and approved the final version of the manuscript.

#### Acknowledgments

The skillful technical assistance of Cornelia Kopp and the opportunity for next-generation sequencing within the Collaborative Research Centre 992 Medical Epigenetics (DFG grant SFB 992) at the Deep Sequencing Facility of the Max Planck Institute of Immunobiology and Epigenetics in Freiburg (Dr. Ulrike Bönisch) is greatly acknowledged. The authors further acknowledge the support of the Freiburg Galaxy Team and Professor Rolf Backofen, Bioinformatics, University of Freiburg, Germany, funded by Collaborative Research Centre 992 Medical Epigenetics (DFG grant SFB 992) and German Federal Ministry of Education and Research (BMBF Grant 031 A538A RBC (de.NBI)). This work was funded by internal grants of Universitäts-Herzzentrum Freiburg • Bad Krozingen and the Institute of Experimental and Clinical Pharmacology and Toxicology of the University Medical Center of Freiburg.

#### **Supplementary Materials**

Supplementary table 1: Differentially expressed long RNAs by edgeR-QLF. Supplementary table 2: Differentially expressed short RNAs by edgeR-QLF. (*Supplementary Materials*)

#### References

- K. C. Vickers, B. T. Palmisano, B. M. Shoucri, R. D. Shamburek, and A. T. Remaley, "MicroRNAs are transported in plasma and delivered to recipient cells by high-density lipoproteins," *Nature Cell Biology*, vol. 13, no. 4, pp. 423–435, 2011.
- [2] O. Gidlöf, M. van der Brug, J. Ohman et al., "Platelets activated during myocardial infarction release functional miRNA, which can be taken up by endothelial cells and regulate ICAM1 expression," *Blood*, vol. 121, no. 19, 2013.
- [3] B. Laffont, A. Corduan, H. Plé et al., "Activated platelets can deliver mRNA regulatory Ago2•microRNA complexes to endothelial cells via microparticles," *Blood*, vol. 122, no. 2, pp. 253–261, 2013.
- [4] A. Risitano, L. M. Beaulieu, O. Vitseva, and J. E. Freedman, "Platelets and platelet-like particles mediate intercellular RNA transfer," *Blood*, vol. 119, no. 26, pp. 6288–6295, 2012.
- [5] A. Zampetaki, S. Kiechl, I. Drozdov et al., "Plasma MicroRNA profiling reveals loss of endothelial MiR-126 and other MicroR-NAs in type 2 diabetes," *Circulation Research*, vol. 107, no. 6, pp. 810–817, 2010.
- [6] D. Kaudewitz, P. Skroblin, L. H. Bender et al., "Association of MicroRNAs and YRNAs with platelet function," *Circulation Research*, vol. 118, no. 3, pp. 420–432, 2016.
- [7] P. Willeit, A. Zampetaki, K. Dudek et al., "Circulating MicroR-NAs as novel biomarkers for platelet activation," *Circulation Research*, vol. 112, no. 4, pp. 595–600, 2013.
- [8] C. Stratz, T. Nührenberg, B. L. Fiebich et al., "Controlled type II diabetes mellitus has no major influence on platelet micro-RNA expression: Results from micro-array profiling in a cohort of 60 patients," *Thrombosis and Haemostasis*, vol. 111, no. 5, pp. 902–911, 2014.
- [9] C. Stratz, T. G. Nührenberg, H. Binder et al., "Micro-array profiling exhibits remarkable intra-individual stability of human platelet micro-RNA," *Thrombosis and Haemostasis*, vol. 107, no. 4, pp. 634–641, 2012.
- [10] D. Kim, G. Pertea, C. Trapnell, H. Pimentel, R. Kelley, and S. L. Salzberg, "TopHat2: accurate alignment of transcriptomes in the presence of insertions, deletions and gene fusions," *Genome Biology*, vol. 14, no. 4, article R36, 2013.
- [11] A. Dobin, C. A. Davis, F. Schlesinger et al., "STAR: ultrafast universal RNA-seq aligner," *Bioinformatics*, vol. 29, no. 1, pp. 15– 21, 2013.
- [12] S. Anders, P. T. Pyl, and W. Huber, "HTSeq—a Python framework to work with high-throughput sequencing data," *Bioinformatics*, vol. 31, no. 2, pp. 166–169, 2015.

- [13] M. I. Love, W. Huber, and S. Anders, "Moderated estimation of fold change and dispersion for RNA-seq data with DESeq2," *Genome Biology*, vol. 15, article 550, 2014.
- [14] H. Li, B. Handsaker, A. Wysoker et al., "The sequence alignment/map format and SAMtools," *Bioinformatics*, vol. 25, no. 16, pp. 2078-2079, 2009.
- [15] A. T. L. Lun, Y. Chen, and G. K. Smyth, "It's DE-licious: A recipe for differential expression analyses of RNA-seq experiments using quasi-likelihood methods in edgeR," *Methods in Molecular Biology*, vol. 1418, pp. 391–416, 2016.
- [16] F. Zambelli, F. Mastropasqua, E. Picardi, A. M. D'Erchia, G. Pesole, and G. Pavesi, "RNentropy: an entropy-based tool for the detection of significant variation of gene expression across multiple RNA-Seq experiments," *Nucleic Acids Research*, vol. 46, no. 8, pp. e46–e46, 2018.
- [17] A. Bisognin, G. Sales, A. Coppe, S. Bortoluzzi, and C. Romualdi, "MAGIA2: From miRNA and genes expression data integrative analysis to microRNA-transcription factor mixed regulatory circuits (2012 update)," *Nucleic Acids Research*, vol. 40, no. 1, pp. W13–W21, 2012.
- [18] B. P. Lewis, C. B. Burge, and D. P. Bartel, "Conserved seed pairing, often flanked by adenosines, indicates that thousands of human genes are microRNA targets," *Cell*, vol. 120, no. 1, pp. 15–20, 2005.
- [19] A. Krek, D. Grün, M. N. Poy et al., "Combinatorial microRNA target predictions," *Nature Genetics*, vol. 37, no. 5, pp. 495–500, 2005.
- [20] K. C. Miranda, T. Huynh, and Y. Tay, "A pattern-based method for the identification of MicroRNA binding sites and their corresponding heteroduplexes," *Cell*, vol. 126, no. 6, pp. 1203– 1217, 2006.
- [21] A. Elgheznawy, L. Shi, J. Hu et al., "Dicer cleavage by calpain determines platelet microRNA levels and function in diabetes," *Circulation Research*, vol. 117, no. 2, pp. 157–165, 2015.
- [22] M. I. Schmidt, B. B. Duncan, A. R. Sharrett et al., "Markers of inflammation and prediction of diabetes mellitus in adults (Atherosclerosis Risk in Communities study): a cohort study," *The Lancet*, vol. 353, no. 9165, pp. 1649–1652, 1999.
- [23] Z. Fejes, S. Póliska, Z. Czimmerer et al., "Hyperglycaemia suppresses microRNA expression in platelets to increase P2RY12 and SELP levels in type 2 diabetes mellitus," *Thrombosis and Haemostasis*, vol. 117, no. 03, pp. 529–542, 2017.
- [24] C. L. Leung, D. Sun, M. Zheng, D. R. Knowles, and R. K. H. Liem, "Microtubule actin cross-linking factor (MACF): A hybrid of dystonin and dystrophin that can interact with the actin and microtubule cytoskeletons," *The Journal of Cell Biology*, vol. 147, no. 6, pp. 1275–1285, 1999.
- [25] J. E. Italiano Jr., W. Bergmeier, S. Tiwari et al., "Mechanisms and implications of platelet discoid shape," *Blood*, vol. 101, no. 12, pp. 4789–4796, 2003.
- [26] J. G. White, "Effects of colchicine and Vinca alkaloids on human platelets. I. Influence on platelet microtubules and contractile function.," *The American Journal of Pathology*, vol. 53, no. 2, pp. 281–291, 1968.
- [27] R. Flaumenhaft, J. R. Dilks, N. Rozenvayn, R. A. Monahan-Earley, D. Feng, and A. M. Dvorak, "The actin cytoskeleton differentially regulates platelet α-granule and dense-granule secretion," *Blood*, vol. 105, no. 10, pp. 3879–3887, 2005.
- [28] H. Kashiwagi, M. A. Schwartz, M. Eigenthaler, K. Davis, M. H. Ginsberg, and S. J. Shattil, "Affinity Modulation of Platelet Integrin α," *The Journal of Cell Biology*, vol. 137, no. 6, pp. 1433– 1443, 1997.

- [29] C. Angénieux, B. Maitre, A. Eckly, F. Lanza, C. Gachet, and H. De La Salle, "Time-dependent decay of mRNA and ribosomal RNA during platelet aging and its correlation with translation activity," *PLoS ONE*, vol. 11, no. 1, 2016.
- [30] L. C. Edelstein, L. M. Simon, R. T. Montoya et al., "Racial differences in human platelet PAR4 reactivity reflect expression of *PCTP* and *miR-376c*," *Nature Medicine*, vol. 19, no. 12, pp. 1609–1616, 2013.
- [31] V. Kameswaran, N. C. Bramswig, L. B. McKenna et al., "Epigenetic regulation of the DLK1-MEG3 MicroRNA cluster in human type 2 diabetic islets," *Cell Metabolism*, vol. 19, no. 1, pp. 135–145, 2014.
- [32] A. A. Alhasan, O. G. Izuogu, H. H. Al-Balool et al., "Circular RNA enrichment in platelets is a signature of transcriptome degradation," *Blood*, vol. 127, no. 9, pp. e1–e11, 2016.
- [33] O. G. Izuogu, A. A. Alhasan, H. M. Alafghani, M. Santibanez-Koref, D. J. Elliot, and M. S. Jackson, "PTESFinder: A computational method to identify post-transcriptional exon shuffling (PTES) events," *BMC Bioinformatics*, vol. 17, no. 1, 2016.
- [34] U. S. Tantry, L. Bonello, D. Aradi et al., "Consensus and update on the definition of on-treatment platelet reactivity to adenosine diphosphate associated with ischemia and bleeding," *Journal of the American College of Cardiology*, vol. 62, no. 24, pp. 2261–2273, 2013.