

1 **Transfer RNA fragments replace microRNA regulators of the**
2 **cholinergic post-stroke immune blockade**

3 Short title: tRF/miR balancers of post-stroke immunity

4

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26

27 **Abstract**

28 Stroke is a leading cause of death and disability. Recovery depends on balance between
29 inflammatory response and immune suppression, which can be CNS-protective but may worsen
30 prognosis by increasing patients' susceptibility to infections. Peripheral cholinergic blockade of
31 immune reactions fine-tunes this immune response, but its molecular regulators are unknown.
32 Therefore, we sought small RNA balancers of the cholinergic anti-inflammatory pathway in
33 peripheral blood from ischemic stroke patients. Using RNA-sequencing and RT-qPCR, we
34 discovered in patients' blood on day 2 after stroke a "change of guards" reflected in massive
35 decreases in microRNAs (miRs) and increases in transfer RNA fragments (tRFs) targeting
36 cholinergic transcripts. Electrophoresis-based size-selection followed by RT-qPCR validated the
37 top 6 upregulated tRFs in a separate cohort of stroke patients, and independent small RNA-
38 sequencing datasets presented post-stroke enriched tRFs as originating from lymphocytes and
39 monocytes. In these immune compartments, we found CD14+ monocytes to express the highest
40 amounts of cholinergic transcripts. In-depth analysis of CD14+ regulatory circuits revealed
41 minimally overlapping subsets of transcription factors carrying complementary motifs to miRs or
42 tRFs, indicating different roles for the stroke-perturbed members of these small RNA species.
43 Furthermore, LPS-stimulated murine RAW264.7 cells presented dexamethasone-suppressible
44 upregulation of the top 6 tRFs identified in human patients, indicating an evolutionarily conserved
45 and pharmaceutically treatable tRF response to inflammatory cues. Our findings identify tRF/miR
46 subgroups which may co-modulate the homeostatic response to stroke in patients' blood and
47 open novel venues for establishing RNA-targeted concepts for post-stroke diagnosis and
48 therapeutics.

49

50 **Keywords**

51 acetylcholine; ischemic stroke; microRNA; immunology; transfer RNA fragment

52

53 **Introduction**

54

55 The increasing incidence of ischemic stroke is a global burden of growing dimensions, accounting
56 for 12 of the 15 million strokes annually, and leaving most of the surviving patients permanently
57 disabled(1). The immune system is one of the main players in the pathophysiology of stroke.
58 Brain injury dampens immune functions in the periphery, which limits the inflammatory response
59 and infiltration of immune cells into the CNS and may pose a neuroprotective mechanism in
60 stroke patients. Importantly, this systemic immunosuppression simultaneously increases the risk
61 of infectious complications(2), e.g. by inducing lymphocyte apoptosis and decreasing production
62 of pro-inflammatory cytokines (lymphocytic $IFN\gamma$ and monocytic $TNF\alpha$)(3). Therefore, post-stroke
63 recovery largely depends on a delicate balance between inflammation, which exacerbates the
64 severity of symptoms, and the post-stroke suppression of immune functions, which increases the
65 susceptibility to infections(3). This involves incompletely understood molecular regulators of
66 cholinergic and sympathetic signaling and the hypothalamus-pituitary-adrenal gland axis (HPA).
67 Among other processes, brain injury leads to activation of the vagus nerve, which mediates anti-
68 inflammatory signaling through the cholinergic efferent fibers and the noradrenergic splenic
69 nerve. Binding of acetylcholine (ACh) to the nicotinic alpha 7 receptors on
70 monocytes/macrophages decreases the production of proinflammatory cytokines(4) in a manner
71 susceptible to suppression by microRNA (miR) regulators of cholinergic signaling, such as miR-
72 132(5). We hypothesized that this and other small RNA fine-tuners of innate immune responses,
73 including miRs and the recently re-discovered transfer RNA (tRNA) fragments (tRFs) may
74 contribute to regulation of post-stroke processes.

75

76 Both miRs and tRFs may control entire biological pathways, such that their balanced
77 orchestration could modulate brain-induced systemic immune functioning. miRs are small non-
78 coding RNAs whose expression requires transcription yet can be rapidly induced, enabling

79 degradation and/or translational suppression of target genes carrying a complementary motif.
80 One miR may suppress the expression of many targets involved in the same biological pathway,
81 and many miRs may co-target the same transcripts, enabling cooperative suppression. Hence,
82 miR regulators of ACh signaling may shape the role of ACh in both cognitive function and
83 systemic inflammation(5, 6).

84 Recent reports highlight tRNA as another major source of small noncoding RNA(7), including
85 tRNA halves (tiRNAs) and smaller tRNA fragments - tRFs. tiRNAs are created by angiogenin
86 cleavage at the anticodon loop(8) raising the possibility that the post-stroke angiogenin increase
87 might change their levels(9). Among other functions(10, 11), smaller fragments derived from the
88 3'- or 5'-end of tRNA (3'-tRF/5'-tRF) or internal tRNA parts (i-tRF) may incorporate into Argonaute
89 (Ago) protein complexes and act like miRs to suppress their targets(12). Differential expression of
90 tRFs was reported under hypoxia, oxidative stress, ischemic reperfusion(8, 13) and in
91 epilepsy(14), which are all involved in ischemic stroke complications. tRFs may be generated via
92 enzymatic degradation of tRNA, independent of de-novo transcription, which implies that tRF
93 levels may be modulated more rapidly than miR levels. However, whether brain-body
94 communication and immune suppression after ischemic stroke in human patients involves blood
95 tRF changes has not yet been studied.

96
97 Taking into consideration that the cholinergic system is one of the controllers of immune
98 functions, we looked for changes in the levels of miR and tRF regulators, with a specific focus on
99 those which may control the ACh-mediated suppression of post-stroke immune functions. We
100 performed small and long RNA-sequencing of whole blood samples collected from stroke patients
101 two days after stroke onset, mined RNA-sequencing datasets of blood cell transcripts and sought
102 potential links between perturbed miRs and tRFs, post-stroke immune responses and the
103 cholinergic anti-inflammatory pathway.

104

105

106 **Results**

107

108 ***A cholinergic-targeting shift from miRs to tRFs in post-stroke patients' blood***

109 To seek post-stroke small RNA regulators of body-brain communication, we first performed small
110 RNA-sequencing of whole blood samples collected on day 2 after ischemic stroke from 33 male
111 patients of the PREDICT cohort (484 participants)(15) and 10 age- and sex-matched controls
112 (Figure 1A; see demographic data in Data file S1). Principal Component Analysis (PCA) of the
113 differentially expressed (DE) small RNAs completely segregated the stroke and control groups
114 (Fig. 1B). The respective direction of change among the two small RNA classes involved a
115 statistically significant decline in miRs and a parallel increase in tRFs, indicating a 'change of
116 guards' from miRs to tRFs. Specifically, 87% of the 143 DE tRFs were upregulated, whereas 63%
117 of the 420 DE miRs were downregulated (Benjamini-Hochberg corrected $p < 0.05$; Fig. 1C, D). Of
118 the 143 DE tRFs, 87 were 3'-tRFs, and 30 of those (all upregulated) were derived from alanine
119 binding tRNA (Supplementary Fig. S1), indicating non-arbitrary fragment generation.

120 Notably, the 420 DE miRs included several miRs known to be perturbed in stroke: miR-532-5p
121 ($\log_{2}FC = -2.27$, $p = 1.81e-33$) (16), miR 148a-3p ($\log_{2}FC = -2.30$, $p = 9.61e-19$) and let-7i-3p
122 ($\log_{2}FC = -1.07$, $p = 4.31e-04$) (17). To test the potential involvement of miRs and tRFs in
123 regulating the cholinergic anti-inflammatory pathway after stroke, we performed targeting analysis
124 of DE miRs and tRFs towards cholinergic transcripts (Supplementary Methods, for a complete list
125 of cholinergic genes see Data File S2) via an in-house integrative database (*miRNeo*)(18)
126 containing comprehensive transcription factor(TF)- and miR-targeting data, complemented by de-
127 novo prediction of tRF-targeting using TargetScan(19). A restrictive approach identified subsets
128 of 131/420 miRs and 64/143 tRFs containing complementary motifs to at least five cholinergic-
129 associated transcripts each (further termed "Cholino-miRs" and "Cholino-tRFs", Fig. 1C, D,
130 Supplementary Fig. S2, full lists in Data files S3 and S4). Permutation targeting analysis showed
131 an enrichment of cholinergic targets for both DE miRs (100 000 permutations, $p = 0.0036$) and
132 DE tRFs (100 000 permutations, $p = 2e-05$). Further indicating non-random generation of these

133 fragments, the tRFs identified in our dataset clustered into oligonucleotide families with high
134 sequence homology via t-distributed stochastic neighbor embedding (t-SNE) (Fig. 1E), including
135 families known to associate with Ago and suppress growth and proliferation via post-
136 transcriptional downregulation in lymphocytes (e.g. tRF-22-WE8SPOX52 from tRNAGly)(20) and
137 metastatic cancer cells (tRF-18-HR0VX6D2 from tRNA^{Leu})(21). This supported our prediction
138 that the concomitant elevation of tRF- and decline of miR-levels in post-stroke blood could
139 contribute to the post-stroke changes in cholinergic signaling pathways.

140 To further challenge our findings, we validated the expression levels of prominently DE tRFs
141 identified by RNA sequencing in a separate cohort of PREDICT patients(15). Standard qPCR
142 techniques cannot distinguish between the full length tRNA molecules and their 3'-tRF cleavage
143 products. Therefore, to experimentally validate tRF changes (Fig. 2A) we implemented an
144 electrophoresis size selection-based strategy followed by cDNA synthesis from the selected small
145 RNAs and RT-qPCR (maximum 25 nt, Fig. 2B). This procedure validated the top six upregulated
146 tRFs identified in RNA-sequencing data (tRF-22-WEKSPM852, tRF-18-8R6546D2, tRF-18-
147 HR0VX6D2, tRF-18-8R6Q46D2, tRF-22-8EKSP1852 and tRF-22-WE8SPOX52; according to
148 count-change, see Supplementary Methods, Fig. 2A and C) and demonstrated significant
149 increases accompanied by higher variability in the blood levels of these tRFs in post-stroke
150 patients compared to controls.

151

152 ***Stroke-regulated whole blood tRFs reflect distinct immune cell compartments***

153 To clarify the distribution of stroke-perturbed tRFs among the immunologically relevant blood cell
154 types, we mined an RNA-sequencing dataset comprising sorted cell populations collected from
155 healthy volunteers: CD4+ T helper cells, CD8+ T cytotoxic cells, CD56+ NK cells, CD19+ B cells,
156 CD14+ monocytes, CD15+ neutrophils, CD235a+ erythrocytes, serum, exosomes, and whole
157 blood (450 samples in total, (22)) (Fig. 3A). Predicting that log-normal distribution of the counts in
158 different samples would point towards biological significance, we categorized all tRFs found in
159 this dataset into present/absent in a specific blood compartment (without introducing a limit for

160 counts, see Methods, Fig. 3B). Two main clusters of specific blood compartments could be
161 identified based on their specific tRF profile: a) monocytes, B-, T- and NK cells; b) neutrophils,
162 whole blood, serum, exosomes and erythrocytes (Supplementary Fig. S3). Further, we
163 distinguished eight tRF sub-clusters, based on the presence/absence of specific tRFs in blood
164 compartments (Fig. 3C), with cluster four comprising molecules expressed specifically in
165 monocytes, B-, T-, and NK-cells, and cluster seven consisting of tRFs expressed only in
166 monocytes.

167 Using the presence/absence measure for analyzing the post-stroke DE tRFs (Fig. 3D,
168 Supplementary Fig. S4 for the top 20 stroke DE tRFs), we detected 77 DE tRFs from the
169 PREDICT dataset as expressed in immune cells (Fig. 3E, a detailed list of tRFs and affiliated
170 clusters in Data File S5), including 10 Cholino-tRFs. Notably, tRFs previously shown to function
171 post-transcriptionally in a miR-like manner (e.g. tRF-22-WE8SPOX52 from tRNAGly(20) and tRF-
172 18-HR0VX6D2 from tRNA^{Leu}, alias miR-1280(21)) segregated into whole blood, monocyte, T-,
173 B-, and NK-cell compartments rather than into erythrocyte, serum or exosome compartments.
174 Thus, the post-stroke modified tRFs may be functionally involved in regulating the leukocytic post-
175 stroke response.

176

177 ***CD14+ monocytes show highest transcriptional activation towards cholinergic genes***

178 The enrichment of DE Cholino-miRs and Cholino-tRFs identified in the PREDICT dataset and the
179 contribution of the cholinergic anti-inflammatory pathway to peripheral immunosuppression called
180 for pinpointing the immune compartment(s) in which these small RNAs might affect post-stroke
181 immune suppression. Analysis of long RNA regulatory circuits specific to blood-borne leukocytes
182 (Fig. 4A)(23) identified CD14+ monocytes as the main cell type expressing cholinergic core and
183 receptor genes (Data File S2, Fig. 4B). To challenge the relevance of this effect for leukocyte
184 features, we performed long RNA-sequencing in blood samples from 20 stroke patients from the
185 PREDICT study and 4 controls. This showed 204 upregulated and 490 downregulated long RNA
186 transcripts. Gene ontology (GO) enrichment analyses of the most implicated genes yielded highly

187 specific terms relevant to innate immunity, vascular processes, and cholinergic links (Fig. 4C, list
188 of all significant terms in Supplementary Table 1). More specifically, terms linked to innate
189 immune processes in post-stroke blood involved *responses to LPS mediated by interferons and*
190 *other cytokines* (Figure 4C, left-hand side); vascular processes comprised *platelet activation and*
191 *degranulation, control of cell-cell adhesion, and regulation of angiogenesis* (Figure 4C, right-hand
192 side). Intriguingly, differentially regulated genes also showed involvement in *response to*
193 *organophosphorus*, which are known acetylcholinesterase (AChE) inhibitors, implying a
194 cholinergic participation.

195

196 ***tRFs may suppress inflammation and cholinergic-associated transcription factors alone or***
197 ***in cooperation with miRs***

198 Cellular responses to different stimuli are coordinated by cell type-specific transcriptional
199 regulatory circuits. To facilitate understanding of the role of miRs and tRFs in regulating the
200 transcriptional state of CD14⁺ monocytes after stroke, we generated a monocyte-specific
201 transcriptional interaction network of small RNAs targeting transcription factors (via *miRNeo*)(18),
202 combined with differential expression of long and small RNAs from the PREDICT cohort (Fig. 5A).
203 Notably, the force-directed network of all TFs active in CD14⁺ monocytes self-segregated to form
204 two distinct clusters of TFs which were primarily targeted either by miRs or tRFs, including
205 numerous TFs DE in stroke patient blood (Fig. 5B). The gradually divergent targeting of these
206 TFs by miRs and/or tRFs implied largely separate domains of regulation by either small RNA
207 species (Fig. 5C). Among the implied TFs are proteins known for their influence on cholinergic
208 genes as well as their involvement in inflammation, such as STAT1 or KLF4(24, 25). Intriguingly,
209 8 DE TFs were not predicted to be targeted by any miR or tRF present in CD14⁺ monocytes (Fig.
210 5D).

211 Next, we aimed to test if the stroke-induced tRFs are involved in the inflammatory response of
212 monocytes and macrophages. For this purpose, we subjected murine RAW264.7 cells to LPS-
213 stimulation with or without dexamethasone suppression of their inflammatory reactions (Fig. 5E).

214 By 18h after LPS stimulation, RT-qPCR analysis based on size-selection (for <50 nt fragments)
215 detected pronounced upregulation of the top 6 post-stroke upregulated tRFs (Fig. 5F). Moreover,
216 dexamethasone suppression of the LPS response prevented increase of those tRFs, along with a
217 diminished inflammatory response (Supplementary Fig. S5). Predicted targets of these molecules
218 comprise members of mitogen-activated protein kinases (MAPK) and tumor necrosis factor
219 receptor-associated factors (TRAF) (see Data File S6), further pointing towards their regulatory
220 role in response to inflammatory stimuli. Together, these findings demonstrate evolutionarily
221 conserved increases of stroke-induced tRFs under pro-inflammatory insults and dexamethasone-
222 mediated suppression of their generation in mononuclear immune cells.

223

224

225 **Discussion**

226

227 We discovered a stroke-induced decline of miRNAs and concomitant elevation of tRFs in whole
228 blood, and demonstrated that this shift may modulate the post-stroke cholinergic blockade of
229 immune function. To validate our RNA-seq findings of tRNA fragments in a way that circumvents
230 the ambiguous detection of full-length tRNA, we developed and used a size selection-based RT-
231 qPCR test in an independent cohort of patients. Mining transcriptomic datasets identified CD14+
232 monocytes as highly involved in the cholinergic control of immunity, demonstrated that the stroke-
233 induced tRFs may target specific monocytic TFs, and showed that at least some of those tRFs
234 may actively control processes linked to inflammatory responses. Moreover, several of the stroke-
235 induced tRFs were also induced in LPS-exposed murine macrophages in a dexamethasone-
236 suppressible manner, supporting the notion that the elevation of tRFs is an evolutionarily
237 conserved mechanism. This novel concept for the fine-tuning of post-stroke immune response
238 opens new venues for post-stroke diagnostics and therapeutics.

239

240 The cholinergic anti-inflammatory reflex plays a substantial role in regulating peripheral immune
241 responses after CNS injury, along with the HPA axis and sympathetic signaling(3). Excessive
242 cholinergic responses suppress pulmonary innate immunity, including macrophage and alveolar
243 epithelial cell responses and facilitate the development of pneumonia(26), a major factor of non-
244 recovery(27). However, while reduced AChE activities in post-stroke patients' serum associate
245 with poor survival(28), stroke-induced immunosuppression may be brain-protective(2), calling for
246 caution when considering therapeutic boosting of immune reaction in the periphery to limit
247 infections. Therefore, an in-depth understanding and characterization of the molecular regulators
248 of immune responses and the cholinergic pathway after CNS-injury is of utmost importance.

249 Translational control is important for regulating stress and immunity(29), and tRFs may be rapidly
250 produced by regulated nuclease cleavage of pre-existing tRNAs in a "burst-like" fashion, which
251 makes them particularly appropriate for handling acute situations. Recent reports demonstrate
252 production of 3'-tRFs by specific nucleases, and 3'- and 5'-tRNA^{Leu} fragments were shown to
253 regulate T cell activation(30). Furthermore, tRFs can perform different molecular roles, including
254 Ago-mediated suppression of target genes carrying complementary sequence motifs(12). At least
255 two of the stroke-induced tRFs upregulated after LPS stimulation show miR-like function: tRF-22-
256 WE8SPOX52 regulates B cell growth via suppressing the expression of Replication Protein A1
257 (RPA1)(20) and a 17 nucleotides-long fragment of tRF-18-HR0VX6D2 limits cancer cell
258 proliferation by impacting the cholinergic-regulating Notch signaling pathway(21). Interestingly,
259 miR-1260b, identified in our study and by others as perturbed post-stroke(31) differs from tRF-18-
260 HR0VX6D2 by one nucleotide at position 9 and an additional nucleotide at the 3'-end
261 (Supplementary Fig. S6), which indicates that miR-1260b may actually be a tRF(32), and calls for
262 further investigations.

263 Apart from their miR-like activities, tRFs may have other functions. For instance, tRNA^{Leu}-CAG
264 fragments facilitate translation and ribosome biogenesis(11), whereas tRNA^{Gly}, tRNA^{Glu},
265 tRNA^{Asp} and tRNA^{Tyr}-derived tRFs displace RNA-binding proteins leading to mRNA

266 destabilization(10). Therefore, potential functions of stroke-perturbed tRFs other than Ago-
267 mediated suppression of gene expression should be further tested.

268

269 Like tRFs, miRs may exert diverse effects depending on their mRNA target site, yet their primary
270 mode of action is translational repression. We found 420 DE miRs, some of which are known to
271 be influenced by CNS injury, in the blood of post-stroke PREDICT patients. A recent whole blood
272 microarray survey identified 15 miRs, including 11 replicated in our study, to be suppressed within
273 less than 72 hours in intracerebral hemorrhage patients compared to controls(33). All of those
274 miRs target multiple immune-related pathways, including Toll-like receptors, neurokinin/JAK-STAT
275 signaling and natural killer cell-mediated cytotoxicity(33). We conclude that the observed changes
276 in both miRs and tRFs may be functionally relevant for post-stroke immunity and recovery
277 prospects.

278

279 We have found CD14+ monocytes to be the most prominent immune subpopulation in cholinergic
280 responses. The role of monocytes in stroke-associated infections has been already established:
281 post-stroke responses include prolonged monocytosis, deactivation and functional impairment of
282 circulating monocytes/macrophages observed in experimental models(34) and human
283 patients(35). Moreover, stroke leads to over-production of CD14++ CD16- (classical) and
284 CD14++ CD16+ (intermediate) monocytes with simultaneous decrease in CD14+CD16++
285 (nonclassical) monocytes, which correlates with stroke-associated infection(36). The response to
286 stroke in blood cells follows specific kinetics: day 2 post ischemia features an increase in STAT3
287 phosphorylation in monocyte subsets when compared to long-term immune responses, which is a
288 hallmark of innate immune reaction to tissue injury that is also detected in patients after major
289 surgery(37). Conversely, STAT3 signaling is linked to immunosuppression in monocytic myeloid-
290 derived suppressor cells (M-MDSC, CD11b+HLA-DR-CD14+CD15-)(38), but causes immune
291 stimulation in monocytes(37). Therefore, the biological activities of stroke-induced tRFs may also
292 be context-specific.

293

294 Stroke is characterized by an initial inflammatory response followed by immunosuppression
295 facilitated by, among others, the cholinergic anti-inflammatory reflex(3). Therefore, the tRFs and
296 Cholino-tRFs in particular may offer new mechanisms of homeostatic fine-tuning in response to
297 cerebral ischemia. Further, not only the peripheral but also the central immune response at the
298 site of the injury is of great importance for stroke prognosis. Brain injury triggers activation of
299 microglia and infiltration of peripheral immune cells, including monocyte-derived macrophages,
300 which accumulate at the lesion site 3-7 days after stroke(36). Experimental evidence highlights
301 essential roles of these cells in CNS-repair processes and neuronal protection(39, 40), and our
302 own studies indicate small RNA-regulated cross-talk between neuronal and immunological
303 regulation by JAK/STAT-related mechanisms(18). Our current study presents tRFs as potential
304 players in regulation of these neuroinflammatory responses.

305

306 Our targeting predictions indicate that the stroke-induced tRFs and suppressed miRs may both
307 regulate monocytic transcriptional states and imply the existence of exclusive and shared
308 regulatory domains between these two small RNA species. Thus, the post-stroke “change of
309 guards” in the small RNA response may lead to preferential de-repression of miR targets and
310 concomitant repression of tRF targets. The de-repression of miR targets may hence be as pivotal
311 for regulating the initial inflammatory response and subsequent peripheral immunosuppression as
312 the tRF elevation we identified. For example, KLF4, identified as down-regulated in our
313 sequencing dataset, is involved in controlling the macrophage response to LPS(25) and the
314 differentiation of monocytes towards an inflammatory phenotype(41). Therefore, a decrease in
315 miRs targeting this TF may contribute to pro-inflammatory monocytic response observed in the
316 initial phase of stroke. Similarly, MAFB is essential in facilitating the clearance of damage-
317 associated molecular patterns (DAMPs) in the ischemic brain, and, consequentially, limiting the
318 inflammatory response while supporting recovery(42). MAFB de-repression in peripheral immune
319 cells may be a mechanism supporting monocyte infiltration of the brain. Conversely, STAT1 and

320 ATF3 may be preferentially repressed due to their targeting by tRFs. STAT1 is essential in IFN-
321 and IL-6-mediated inflammatory response, and ATF3 is similarly induced by IFNs and contributes
322 to STAT activity via inhibition of STAT-dephosphorylating phosphatases(43, 44). Additionally,
323 ATF3 down-regulates AChE expression during stress(45). Whether these processes contribute to
324 body homeostasis after the traumatic event, or rather to pathologic derailment of immune
325 function, requires detailed kinetic studies of circulating monocytes and brain-infiltrating monocyte-
326 derived macrophages, with simultaneous profiling of short and long transcripts.

327

328 While the specific roles of tRFs in regulating local neuroinflammatory responses and functional
329 modulation of specific peripheral monocyte subsets remain to be elucidated, our findings point
330 towards tRFs/miRs as interchangeable regulators of post-stroke immune responses and potential
331 biomarkers for increased infection risk in these patients. The cumulative role of tRFs and miRs as
332 general post-traumatic mediators of CNS-immune communication thus calls for seeking small
333 RNAs, and tRFs in particular, as involved in other traumatic pathologies such as spinal cord
334 injury, traumatic brain injury, concussion, as well as neuroinflammatory brain diseases.

335

336

337 **Materials and Methods**

338

339 Expanded methods can be found in the online supplement.

340

341 **Clinical cohort**

342 PREDICT was a prospective multi-center study with sites in Germany and Spain
343 (www.clinicaltrials.gov, NCT01079728)(15) that analyzed 484 acute ischemic stroke patients.
344 Patients underwent daily screenings for stroke-associated pneumonia, dysphagia and
345 inflammation markers and their clinical outcome was recorded 3 months post-stroke. To exclude
346 very severe cases of stroke, we only considered for sequencing samples from patients with

347 modified Rankin Scale (mRS) values of 3 and below at discharge from the hospital, leaving
348 n=240 relevant cases. Blood was collected into RNA stabilizing tubes (Tempus Blood RNA tubes,
349 Applied Biosystems™) on each day of hospitalization. Blood samples collected on the 2nd day
350 were subjected to small and long RNA-sequencing, with time from stroke occurrence to blood
351 withdrawal varying between 0.94 to 2.63 days (average: 1.98 days). Blood samples from age-
352 and ethnicity-matched healthy controls were obtained at matched circadian time from donors with
353 ethical approvals from institutional review boards (ZenBio, North Carolina, USA).

354

355 **RNA extraction, quality control and sequencing**

356 RNA was extracted from 3 ml of whole blood of 484 PREDICT patients using Tempus Spin RNA
357 isolation kit (Invitrogen, Thermo Fisher Scientific, Waltham MA, USA). Pre-sequencing
358 Bioanalyzer 6000 (Agilent, Santa Clara CA, USA) tests showed high RNA quality (RIN values 7.9-
359 9.9, median 8.8). Libraries constructed from 600 ng total RNA of 43 samples were subjected to
360 small RNA-sequencing (NEBNext® Multiplex Small RNA library prep set for Illumina, New
361 England Biolabs, Ipswich MA, USA), and 24 of the small RNA-sequenced samples served for
362 PolyA-selected long mRNA sequencing (1000 ng total RNA per sample, TruSeq RNA library
363 preparation kit (Illumina, San Diego CA, USA)). Sequencing (24 or 12 samples per flow cell for
364 small and long RNAs, respectively) was performed on the Illumina NextSeq 500 platform at the
365 Hebrew University's Center for Genomic Technologies.

366

367 **Alignment and count table generation of RNA sequencing reads**

368 Quality control was performed using FastQC, version 0.11.2. Flexbar(46) (with parameters “-q
369 TAIL -qf Sanger -qw 4 -min-read-length 16”)(47) served for adapter trimming and quality based
370 filtering of all raw reads. Long RNA was aligned to the human reference transcriptome
371 (ENSEMBL GRCh38 release 79) using salmon(48) with default parameters. Small RNA was
372 aligned to the miRBase version 21 using miRExpress 2.1.4(49) with default parameters but
373 skipping adapter trimming for miR expression, and to the tRNA transcriptome using the MINTmap

374 pipeline(50) with default parameters for tRF expression (using only reads mapping exclusively to
375 the tRNA space). Raw gene-expression data of small and long RNA sequencing and technical
376 covariates are available via the GEO database (GEO accession number GSE112803).

377

378 **Size selection for tRF quantification**

379 Standard RT-qPCR methods do not allow to distinguish between full length tRNA molecules and
380 3'tRFs. To exclude longer RNA species in the RT-qPCR quantifications, we performed RNA size
381 selection on 15% TBE-Urea-Polyacrylamide gels, selecting only RNA molecules ≤ 25 nucleotides
382 for validations in the clinical cohort and ≤ 50 nucleotides for the assessment of tRF expression in
383 LPS-stimulated RAW 264.7 cells. Detailed description can be found in the Supplementary
384 Methods.

385

386 **Analysis of the presence/absence of specific tRFs in blood compartments**

387 In descriptive analysis of small RNA expression, a threshold (e.g., at least 5 counts in at least
388 80% of samples) is often used to define presence or absence of small RNAs. However, since this
389 definition relies heavily on sequencing depth, and depth can vary widely even in methodically
390 robust sequencing experiments depending on a large number of variables, we defined our own
391 test for descriptive analysis of presence or absence of lowly expressed small RNAs in each of the
392 sample types. Briefly, this definition comprises estimation of a log-normal distribution on the
393 expression profile of the small RNA across all samples in the individual cell types, and a statistical
394 test to refute the null hypothesis that the distribution is in fact log-normal. For each small RNA,
395 the distribution mean and standard deviation of the expression values per cell type were
396 estimated using the *fitdist* function of the R/fitdistrplus package(51). The count distribution was
397 then tested against a log-normal distribution with the estimated mean and standard deviation via
398 the R implementation of the Kolmogorov-Smirnov test, with a cutoff of 0.1. The small RNA was
399 defined as present if the test failed to reject the null hypothesis (see Supplementary Fig. S4 for
400 examples).

401

402 **Cell culture experiments**

403 Murine RAW264.7 cells (ATCC TIB-71) cultured according to standard protocols in Dulbecco's
404 Modified Eagle's Medium supplemented with 10% Fetal Calf Serum, 1% Penicillin-Streptomycin-
405 Amphotericin B and 1% L-Glutamine were collected using cell scraper and stimulated with
406 lipopolysaccharide (LPS from *E.coli*, Sigma Aldrich, St. Louis, USA) following modified protocol
407 by Bartosh et al.(52) . Briefly, 2×10^5 cells were stimulated with 100 ng/ml LPS +/- 0.5 μ M
408 dexamethasone per well (Sigma Aldrich, St. Louis, USA) in 12-well cell culture plates. Cells were
409 collected in Tri-Reagent (Sigma Aldrich, St. Louis, USA) 18h after LPS stimulation and RNA was
410 isolated using miRNeasy kit (Qiagen, Hilden, Germany). For the size selection 1 μ g of total RNA
411 was used and cDNA was synthesized from 500 pg size selected RNA using qScript microRNA
412 cDNA Synthesis Kit (Quanta Biosciences, Beverly MA, USA) and following standard protocol (for
413 further details see Supplementary Methods). Data presented in the manuscript is derived from 3
414 independent experiments (2 of them with dexamethasone treatment) with 2-4 technical replicates
415 in each group.

416

417 **Statistical analysis**

418 Data analysis was performed using R (version 3.4.1). FDR correction was applied whenever
419 applicable, and qPCR data was analyzed using Bio-Rad CFX Maestro software (Bio-Rad, Version
420 4.1.2433.1219) and GraphPad Prism 8.0 (GraphPad Prism Software Inc., San Diego, USA).

421

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435
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- 551
- 552

553 **Figures**

554

555 **Figure 1. Post-stroke co-modified blood tRFs/miRs targeting cholinergic transcripts**
556 **segregate into distinct clusters.** A) Whole blood RNAs were collected on day 2 post-stroke
557 from patients of the PREDICT cohort (NCT01079728)(15) and age-matched controls. B) PCA of
558 DE tRFs/miRs in patients' blood separated stroke and control samples. C) Volcano plot of DE
559 tRFs from stroke patients and controls (horizontal line at adjusted $p = 0.05$) showing upregulation
560 of most DE tRFs. D) Volcano plot of DE miRs shows predominant downregulation in stroke
561 patients compared with controls (horizontal line at adjusted $p = 0.05$). Red dots in C and D reflect
562 Cholino-tRFs and Cholino-miRs, respectively. E) t-SNE visualization of tRF homology based on
563 pairwise alignment scores of sequences of all detected tRNA fragments shows grouped tRFs of
564 several specific amino acid origins.

565

566 **Figure 2. RT-qPCR validation in PREDICT stroke patients of top 6 upregulated tRFs**
567 **following size selection for small RNAs.** A) RNA-sequencing counts normalized to the size of
568 the library (using DESeq2(53)) of the top 6 upregulated tRFs (from left to right). Asterisks indicate
569 adjusted p-values of Wald test via DESeq2, * $p < 0.05$, ** $p < 0.01$, *** $p < 0.001$, shown are box-
570 plots with whiskers minimum to maximum. B) Size selection workflow for validations in a separate
571 sub-group of PREDICT stroke patients ($n=32$) using the same control group ($n=10$); C) RT-qPCR
572 validations using normalized expression (hsa-miR-30d-5p, hsa-let7d-5p, hsa-miR-106b-3p and
573 hsa-miR-3615 served as house-keeping genes, see Supplementary Methods), relative to the
574 control group (line at mean normalized expression for the control group =1) confirmed
575 upregulation of top 6 DE tRFs identified in RNA-sequencing, ANOVA with Tukey post-hoc, * $p <$
576 0.05 , ** $p < 0.01$, *** $p < 0.001$, box-plots with whiskers minimum to maximum.

577

578 **Figure 3. Identifying blood tRFs DE post-stroke in specific cellular compartments.** A)

579 Analysis of RNA-sequencing datasets from T lymphocytes (CD4+ T helper cells and CD8+ T
580 cytotoxic cells), B lymphocytes (CD19+), NK cells (CD56+), monocytes (CD14+), neutrophils
581 (CD15+) erythrocytes (CD235a+), serum, exosomes and whole blood(22) yielded a blood tRF
582 profile. B) Definition of presence/absence of small RNAs in these blood compartments via
583 statistical assertion of log-normal count distribution (Values between 0 and 1, closer to 1:
584 present). C) Detailed analysis of identified tRFs found 8 sub-clusters based on cell types
585 expressing specific molecules. D) t-SNE of all found tRFs represented by grey dots, DE tRFs
586 identified in the PREDICT study are marked with cluster-specific color. E) t-SNE of all tRFs found,
587 Cholino-tRFs identified in the PREDICT study are marked with cluster-specific color.

588

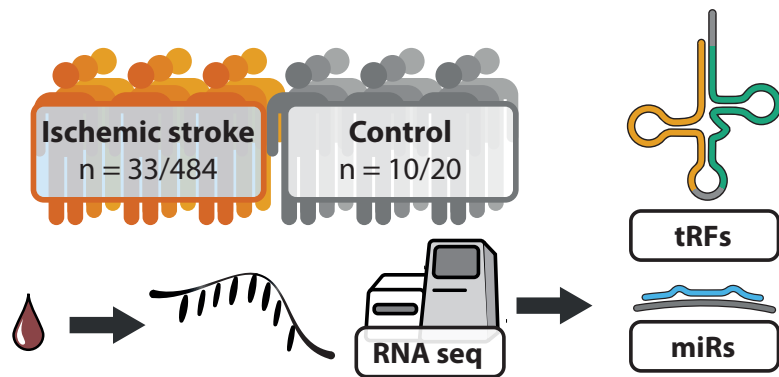
589 **Figure 4. Monocytes express most cholinergic core and receptor genes** A) Shown are web-
590 deposited cell type-specific long RNA profiles(23), tissue-matched to reflect T lymphocytes (CD4+
591 T-helper cells and CD8+ T-cytotoxic cells), B lymphocytes (CD19+), NK cells (CD56+),
592 monocytes (CD14+), and neutrophils (CD15+). B) t-SNE visualization of 15032 genes on the
593 basis of their expression in blood-borne immune cells extrapolated from transcriptional activities
594 in regulatory circuits(23). Genes are colored by the cell type in which their expression was
595 highest. Cholinergic core and receptor genes were mainly found in the CD14+ monocytic
596 compartment. C) Enrichment of post-stroke DE genes ($\log_2\text{FoldChange} > 1.4$) in circulation- and
597 immunity-related pathways, presented as t-SNE of GO terms by their shared genes (see
598 Supplementary Methods); color denotes t-SNE cluster, size denotes number of significant genes
599 in term; deeper color indicates lower enrichment p-value (all p-values < 0.05). Distance between
600 terms indicates the number of shared genes between the GO terms, closer meaning more shared
601 genes.

602

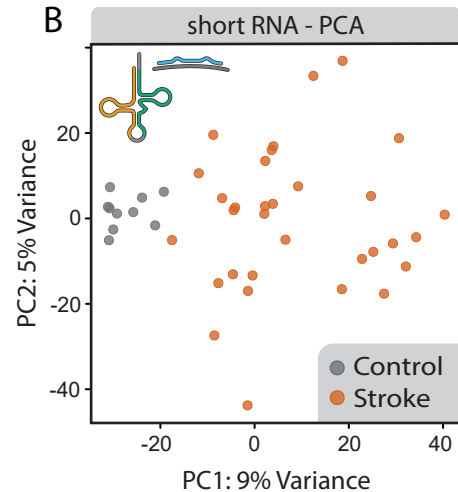
603 **Figure 5. CD14+ TF regulatory circuits reveal dichotomous miRs and tRfs influence, and**
604 **murine RAW 264.7 cells show dexamethasone-suppressible tRF elevation under LPS.** A)

605 Small RNA targeting of TFs active in CD14+ cells was analyzed using an in-house database(18).
606 B) Force-directed network of all TFs active in CD14+ monocytes self-segregates to form largely
607 distinct TF clusters targeted by DE tRFs and miRs in stroke patients' blood. Yellow = TF, red = TF
608 DE in stroke patients' blood, green = miR, purple = tRF. Size of node denotes activity towards
609 targets. C) The top 18 DE TFs in stroke patients' blood present a gradient of targeting by miRs
610 and/or tRFs (left = 100% miR targeting, right = 100% tRF targeting; value shown as "tRF fraction
611 – 0.5" to center on 50/50 regulation by miRs and tRFs). Size of points and color denote absolute
612 count-change and direction of differential regulation, respectively. "C" marks TFs targeting
613 cholinergic core or receptor genes. D) Eight DE TFs present in CD14+ monocytes are predicted
614 not to be targeted by small RNAs. E) LPS stimulation of RAW264.7 murine macrophage cells
615 induced clear morphology changes within 18h. Extracted RNA was subjected to size selection
616 and cDNA synthesized from the <50 nt fraction alone. Scale bar = 100µm F) LPS-stimulated
617 RAW264.7 cells show dexamethasone-suppressible elevated levels of post-stroke induced tRFs.
618 Normalized RT-qPCR values (using mmu-miR-30d-5p, mmu-let7d-5p as house-keeping genes,
619 Supplementary Methods), compared to unstimulated controls. Each dot represents 2-4 technical
620 replicates, ANOVA with Tukey post-hoc, * p < 0.05, ** p < 0.01, *** p < 0.001, bar graphs +/-
621 standard deviation (SD).

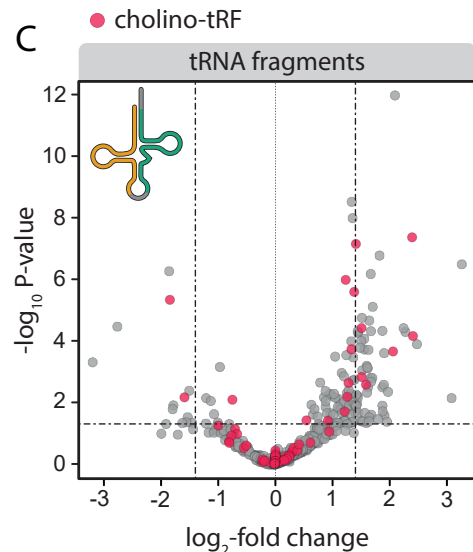
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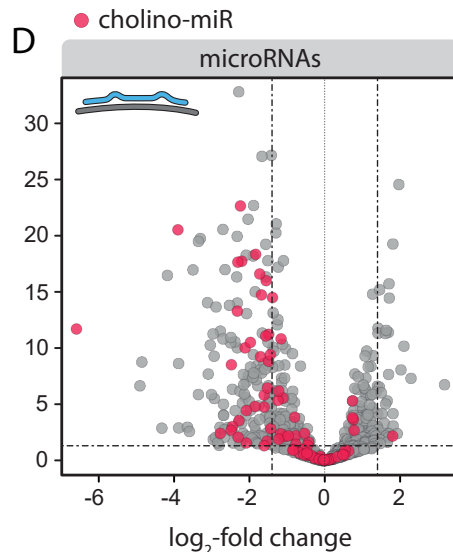
B



C



D



E

