

RESEARCH PAPER

Biomarkers predict outcome in Charcot-Marie-Tooth disease 1A

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ABSTRACT

Background Charcot-Marie-Tooth disease type 1A (CMT1A) is the most common inherited neuropathy, a debilitating disease without known cure. Among patients with CMT1A, disease manifestation, progression and severity are strikingly variable, which poses major challenges for the development of new therapies. Hence, there is a strong need for sensitive outcome measures such as disease and progression biomarkers, which would add powerful tools to monitor therapeutic effects in CMT1A.

Methods We established a pan-European and American consortium comprising nine clinical centres including 311 patients with CMT1A in total. From all patients, the CMT neuropathy score and secondary outcome measures were obtained and a skin biopsy collected. In order to assess and validate disease severity and progression biomarkers, we performed qPCR on a set of 16 animal model-derived potential biomarkers in skin biopsy mRNA extracts.

Results In 266 patients with CMT1A, a cluster of eight cutaneous transcripts differentiates disease severity with a sensitivity and specificity of 90% and 76.1%, respectively. In an additional cohort of 45 patients with CMT1A, from whom a second skin biopsy was taken after 2–3 years, the cutaneous mRNA expression of GSTT2, CTSA, PPARG, CDA, ENPP1 and NRG1-lis changing over time and correlates with disease progression.

Conclusions In summary, we provide evidence that cutaneous transcripts in patients with CMT1A serve as disease severity and progression biomarkers and, if implemented into clinical trials, they could markedly accelerate the development of a therapy for CMT1A.

INTRODUCTION

Charcot-Marie-Tooth (CMT) diseases or hereditary motor and sensory neuropathies comprise a family of the most common inherited disorders of the peripheral nervous system with a prevalence of up to 1 in 1214.^{1,2} So far, >1000 different mutations have been discovered in >90 genes linked to hereditary neuropathy.³ In >60% of all cases, the genetic defect underlying CMT is a duplication of

the gene encoding the 22 kDa peripheral myelin protein (PMP22).^{4–9} Affected individuals develop a slowly progressive demyelinating neuropathy with distally pronounced muscle atrophy and sensory impairment, resulting in steppage gait, altered deep tendon reflexes and skeletal deformities (eg, pes cavus).¹⁰ CMT disease type 1A (CMT1A) typically manifests within the first two decades of life, and walking disabilities, foot deformities and electrophysiological abnormalities in most patients are present already in childhood.^{11,12} However, CMT1A disease onset, progression and severity are strikingly variable within families and even among monozygotic twins.^{13,14} Despite several promising trials in animal models, no causal treatment is available for any form of CMT yet.^{15,16} Neither trials of exercise and orthosis, nor pharmacological approaches with ganglioside, creatine and, more recently, oral administration of ascorbic acid showed beneficial effects in patients with CMT1A.^{17–21} For ascorbic acid, it was suggested that a treatment effect may have been missed because of an unexpectedly slow disease progression.²¹ Hence, insensitive outcome measures harbour the risk to mask therapeutic effects in clinical trials.^{21,22} Improvement of outcome measures is therefore essential to facilitate the development of a treatment for CMT1A disease.^{23,24} To date, only physical and electrophysiological examinations are available to determine disease severity in patients with CMT. The CMT neuropathy score (CMTNS) is a valid and reliable nine-item composite scale taking into account sensory and motor symptoms.²⁵ In order to standardise patient assessment and to improve the scale's sensitivity to change, a novel version of the CMTNS has been proposed.^{23,26} Surrogate biomarkers of disease severity and progression would add powerful tools to monitor therapeutic effects in clinical trials.^{15,21,27–29} Recently, non-invasive MRI study has shown muscle water changes and intramuscular fat accumulation in the lower limbs to be correlated to clinical impairment.^{30,31} While promising as a surrogate outcome measure for clinical trials, this approach is technically demanding and has so far only been tested in proof of principle in a small group of patients suffering from CMT



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and various other neuromuscular disorders. Molecular analysis of body fluids or tissue biopsies is a promising strategy for the development of clinically more practical surrogate biomarkers for disease severity in CMT1A.

The development of valid biomarker assays, however, requires a fit-for-purpose approach^{32,33} that can be separated into four iterative phases: (1) the prevalidation phase defines the intended purpose of the biomarker, considering preanalytical variables and bioanalytical method feasibility. (2) The exploratory validation phase assesses the basic assay performance. (3) The advanced validation phase characterises the formal performance of the assay with regard to its intended use. (4) The in-study validation phase that ensures that the assay method performs robustly across studies according to predefined specification and facilitates the establishment of definitive acceptance criteria.^{32,33}

Within an exploratory prevalidation phase (fit-for purpose step 1), employing a *Pmp22* transgenic rat model for CMT1A (CMT rat³⁴), we could previously show that transcriptional profiling in skin biopsies can be used to identify prognostic and disease severity biomarkers which correlate with clinical impairment.¹⁶ Importantly, in a translational proof-of-concept approach, we could validate rodent cutaneous mRNA biomarkers in skin biopsies from 46 patients with CMT1A.¹⁶ Building on these preliminary results, here we aimed at the exploratory validation (fit-for-purpose step 2) of 16 CMT rat-derived potential transcriptional biomarkers within a large pan-European and American consortium providing skin biopsies from 266 clinically well-characterised genetically proven patients with CMT1A. Next to the assessment of clinical impairment, the ability to measure disease progression is an indispensable prerequisite for the successful conduction of therapeutic trials. Given the poor detectability of clinical impairment over time,²¹ disease progression biomarkers could markedly accelerate the development of a therapy. Hence, we aimed at the advanced validation phase of potential biomarkers with regard to disease progression (fit-for-purpose step 3) and collected 45 paired skin biopsies from patients with CMT1A that were reassessed after 2–3 years.

METHODS

Patient recruitment

All clinical centres involved obtained prior ethical approval of their local institutional review board and their respective regional ethics committee. An overall description of the patient recruitment and a detailed listing of all assessed clinical parameters and outcome measures can be found in the (online supplementary material 1).

Skin biopsy

Skin biopsies with a diameter of 3 mm were taken from the proximal, medial part of the index finger of the non-dominant hand. Sterile disposables of 3 mm diameter, 'Biopsy Punch' by Stiefel were used. The skin biopsy was split into two equal parts, one placed in 4% paraformaldehyde in 1 phosphate-buffered saline, the other part was stored in RNAlater (Ambion).

RNA isolation, precipitation and cDNA synthesis

For the gene expression measurements, the total RNA was extracted from one half of the skin biopsies (in RNA later). The sample was homogenised in RLT buffer (Qiagen) using a rotor stator and processed using the Qiagen manufacture protocol for non-fatty tissue (RNeasy Mini Kit, Qiagen). Before cDNA synthesis, the RNA was precipitated by adding 0.5 volumes ammonium acetate (7M, RNA grade), 2 µL pellet paint (EMD

Millipore) and 2.5 volumes pure ethanol (RNA grade) and resuspended in RNase free. The preparation and processing of all samples were performed in parallel by the same person using the same protocol and the quality of RNA was verified by integrity check (Agilent). Only samples with a RNA integrity number >7 were used for further analyses. The cDNA synthesis was carried out using the Superscript III RT kit (Invitrogen). To reversely transcribe mRNA, we used dT Primer (0.6 µM) and random nonamer primers (N9 Primers 120 µM) in a single PCR step. The obtained cDNA was diluted 1 to 10 before qPCR analyses.

Real-time semiquantitative PCR with TaqMan and SYBRGreen

The qPCR was operated in the LightCycler 480 Systems (384-well format, Roche Applied Science) using 2 µL of diluted cDNA from skin biopsies in each reaction. For all genes except *NRG1-I*, TaqMan qPCR assays were performed using the manufacturer's protocol (TaqMan Universal PCR Master Mix, Applied Biosystems). Intron spanning primers (0.9 µM, each) and FAM-TAMRA tagged probes (0.25 µM) were designed by Microsynth (Switzerland) for each assay (see sequences in (online supplementary material 2)). For *NRG1-I*, no primer probe set could be designed and we designed primers amplifying the type I specific 5' region of the *NRG1* gene using Primer Express Software V.1.65 (Applied Biosystems, see sequences in (online supplementary material 2)). For *NRG1-I*, we performed qPCR using the SYBRGreen PCR master mix (Applied Biosystems).

For all qPCR assays, the reaction mix was prepared to the final volume of 10 µL and all reactions were performed in four replicates. Due to the high number of samples, six 384-well plates for each assay were used and the plates were calibrated to each other by analysis of three calibrator samples on each plate for each gene. For quantification of expression levels, the threshold cycles (Ct) for each gene of interest was normalised against two stable housekeeping genes (*B2M* and *RPLP0*).

Data analysis and statistics

Four technical replicates were available for each qPCR measurement which were summarised by their median for further analysis. The difference to the mean expression of the two housekeeper genes *B2M* and *RPLP0* was calculated per patient. To remain on the Ct scale, for each plate the median of all housekeeper gene measurements was added. Three reference patients were quantified on each qPCR plate and their mean expression was subtracted to normalise for plate effects. Again, to remain on Ct scale, the median of all reference patient measurements was added. The resulting normalised expression data are $\Delta\Delta Ct$ values rescaled to a Ct comparable scale.

The secondary clinical parameters were transformed into z scores using mean and SD of healthy controls of corresponding age, class and gender. The z scores were further categorised into one of the levels normal, very mild, mild, moderate and severe (0, 1, 2, 3, 4).

Cohort-specific mean and SD were calculated for age, body mass index (BMI), all primary and secondary clinical parameters as well as different CMT scores.

The normalised Ct values from the PCR experiment were used for correlation-based hierarchical clustering, where patients were clustered on the correlation of their expression profiles across the potential biomarkers. Similarly, the potential biomarkers were clustered based on the correlation of the expression profiles across patients, where both, positive and negative correlation, were interpreted as small distance. Additionally, the potential biomarkers were clustered by their pairwise correlations.

Supervised principal component analysis (PCA) was performed on the PCR data. To account for potential centre and plate effects, linear mixed effects models were fit to the expression data controlling for CMTNS and estimating coefficients for the genes, the plates and the centres. The data were then corrected by the estimated plate and centre effects. CMTNS values were genewise regressed onto the expression values and the expression values were scaled by the resulting coefficient so that the rescaled expression values were in CMTNS unit scale. A canonical PCA was performed on these data. Patients with CMT1A were classified as mild (CMTNS <10), moderate (CMTNS between 10 and 20) or severe (CMTNS >20). A support vector machine (SVM) classifier separating mild and severe CMT1A cases based solely on gene expression profiles was trained and evaluated using a 10 times repeated 10-fold cross-validation.

The association between CMT scores, as well as primary and secondary subscores with the expression of the potential biomarkers, was assessed by fitting linear models to the expression values with the score as predictor while controlling for centre, plate, age, gender and BMI.

Using the data from patients with measurements in two points in time, linear mixed effect models were fit to the CMTNS quantifying the progression effect while controlling for BMI, gender and age effects. Similar mixed effects repeated measures regression models were fit per gene to the expression data, where the effect of time was assessed controlling for BMI, gender, age and centre effects.

A predictive random forest model that classifies patients into progressive patients (CMTNS score increase) and non-progressive patients (CMTNS score does not increase) was trained on the genes with significant change in expression over time. Prediction performance was evaluated in a 10 times repeated 10-fold cross-validation.

Gene set enrichment analysis was performed by the web-based gene list enrichment analysis tool Enrichr (<http://amp.pharm.mssm.edu/Enrichr>³⁵).

RESULTS

In total, 266 plus 45 patients with CMT1A were examined in nine centres around the world (95 from Germany, 56 from the Czech Republic, 50 from Spain, 26 from Belgium, 20 from the UK, 10 from Italy and 9 from France were examined once for the testing of disease severity biomarkers, and 16 from Germany, 14 from Italy and 15 from the USA were examined twice for disease progression biomarkers) (figure 1). From all patients, the CMTNS and related outcome measures were obtained and the patients were characterised according to standardised operating procedures (see section 'Methods') and the patients were further characterised for their gender, age and BMI (table 1). The overall patient cohort (n=266) splits into 155 (58%) female and 111 (42%) male patients with CMT1A, with a mean age of 43 years (SD 14, range 18–80). The CMTNS ranged from 4 to 29 with a median value of 14 (table 1). Importantly, the patient cohorts from the different centres display significant differences in a majority of the obtained parameters (table 1). This intercentre variability may partially be explained by regional differences in the composition of the patient cohort. However, small cohort sizes in some centres or differences in routine clinical examination and assessment of the CMTNS, despite standardised operating procedures, are more likely to account for the intercentre variability. Indeed, for example, the very disparate scores for the dynamometrical secondary outcome measures are in favour of this notion (table 1). Therefore, to avoid disruptive influences of

centre effects in the clinical parameters on the biomarker analyses, the subsequent expression data analysis was controlled for the co-variables centre, age, gender and BMI (table 2).

Distinct cutaneous gene expression profiles reflect disease severity in CMT1A

After clinical examination, the patients were asked to undergo a skin biopsy for biomarker analysis (figure 1A). From the collected skin biopsies, the RNA was purified and qPCR performed for 16 transcripts derived from the previously identified disease severity biomarkers in CMT rats.¹⁶ The selected mRNA biomarkers were readily detectable and display a highly variable expression compared with housekeeping genes (figure 1B). Of note, the expression values among potential biomarkers demonstrate a positive correlation with each other throughout the entire data set (figure 1C). In detail, eight out of 16 tested biomarkers (CDA, CTSA, GRIA1, ENPP1, ANPEP, FN3KRP, GSTT2 and PPARG) show a similar expression pattern among patients with correlation coefficients >0.7 (figure 1A,B). Next, we asked whether we can identify the components in the expression of these eight biomarkers that provide information with respect to CMT1A disease severity. A supervised PCA indeed revealed a principal component 1 (PC1) that can explain 95% of the variance observed in the expression of the biomarker cluster (figure 2A). Importantly, mapping the transcriptional profiles of the available samples to the first two principal components (PC1 and PC2) allows for the separation of the patient cohort according to the CMT status (figure 2B). In detail, PC1 not only separates patients with CMT1A from healthy controls, but also differentiates for disease severity within the patient group (mildly affected: CMTNS <10; moderately affected: CMTNS 10–20; severely affected: CMTNS >20) (figure 2B). Of note, severely affected patients in PC1 cluster in close proximity to healthy controls, whereas mildly affected patients display a more distinct expression profile (figure 2B). An SVM classifier, which was only trained on gene expression profiles of the identified eight biomarkers, separates mildly from severely affected patients (based on the CMTNS) with a sensitivity of 90% and a specificity of 76% (figure 2C). To test to which extent the expression of single genes correlates with disease severity, we next analysed their relation to the CMTNS and individual primary and secondary outcome measures (table 2). As mentioned above, to avoid disruptive influence of centre effects on biomarker correlation, effects were controlled for the co-variables centre, age, gender and BMI. Linear models were fit to the expression data to quantify the influence of each (sub)-score on the expression of each biomarker. The resulting regression coefficients (ie, the factor by which the biomarker expression is changing when the assessed clinical parameter is changing for one unit) show a number of potential biomarkers significantly related to variants of CMTNS and one or more of the subscores (table 2). Importantly, some primary clinical subscores, for example, 'Strength of legs' and 'Vibration', show numerous significant correlations compared with the number of significant hits in the cumulative CMTNS variants (table 2).

The mRNA expression of single genes is changing with disease progression

To assess whether the previously identified biomarkers also harbour information about disease progression, we analysed mRNA expression in 45 sets of paired skin biopsies that were sampled from the very same patients over a 2–3-year time frame (16 from Germany, 14 from Italy and 15 from the

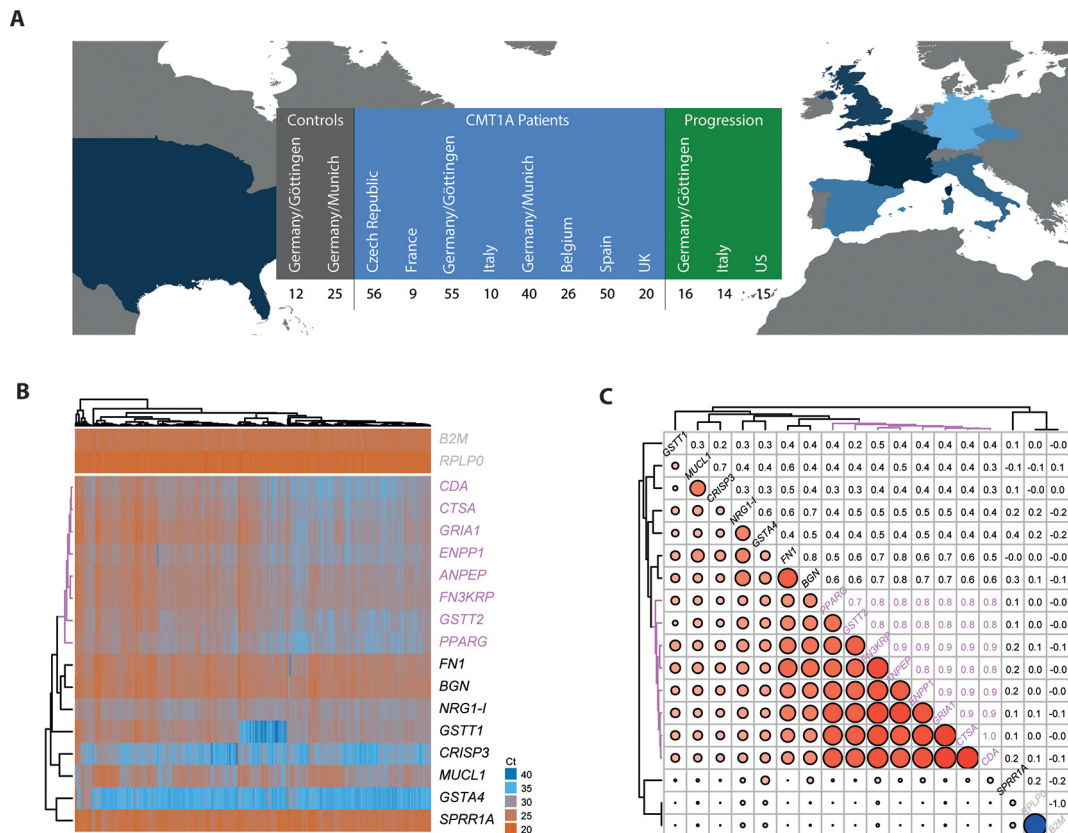


Figure 1 Previously identified potential cutaneous mRNA biomarkers tested in skin biopsies of 266 patients with Charcot-Marie-Tooth disease type 1A (CMT1A). (A) Participating centres in Europe and the USA with number of assessed patients and contributed skin biopsies for biomarker analyses (see [table 2](#) for further details). Next to 266 patients from Europe that were assessed once (in blue) additional 45 patients from Europe and the USA were sampled twice within 2–3 years giving information on the progression (in green). Thirty-seven healthy humans were included as controls from two centres in Germany (grey). (B) Heatmap displaying normalised Ct values from the qPCR analyses of all patients (columns) for all genes (rows). High values (in blue) correspond to low expression while low values (in red) indicate high expression. Both dimensions, patients and genes, are reordered by means of correlation-based hierarchical clustering to group by expression profile similarities. The dendrogram on the top shows the clustering of patients, the dendrogram on the left shows the clustering of the genes. The biggest cluster of eight similarly regulated genes is highlighted in magenta. The two rows at the top show the two housekeeper genes for reference. (C) Correlation of the gene expression profiles across the patients. Both, rows and columns show genes, the upper triangles show the numerical values of the correlation coefficients, the lower triangle visualises the correlation coefficients where blue coloured circles represent negative correlation and red-coloured circles represent positive correlation. The size of the circles indicates the strength of correlation. The genes are clustered by their correlation profile. Also, 8 out of 16 genes cluster with correlation coefficients >0.7 (highlighted in magenta). Most notably, there is no negative correlation among the genes, except for the perfectly negatively correlated housekeeper genes.

USA; [figure 1A](#)). In total, gene expression values were available for two time points for 45 patients, clinical data were available for 46 patients, with 30 patients with CMT1A (mildly affected: n=3, moderately affected: n=23, severely affected: n=4) having both clinical and gene expression values. We could confirm the low mean change in CMTNS of about ~0.7 points per year in our cohort of 46 patients with CMT1A ([figure 3A,B](#)), a value that matches previously published data.³⁶ Mixed model linear regression on repeated measures analysis shows a highly significant effect of ‘time’ (ie, disease progression) on the CMTNS, controlling for gender, age and BMI, pointing to clinical detectability of disease progression in a 2–3-year time frame ([figure 3C](#)). In 45 skin biopsies from patients with CMT1A, we found six genes with significantly increased expression over time: *GSTT2*, *CTSA*, *PPARG*, *CDA*, *ENPP1*, *NRG1-I* ([figure 4A](#)). The extent of increase in gene expression per year ranges from 2^{0.28}=1.2-fold for *NRG1-I* to 2^{0.66}=1.6-fold for *CTSA* ([figure 4B](#)). A predictive random forest model was trained on these six genes. The 30 patients with CMT1A from whom both clinical and biomarker data were available were classified into progressive

(CMTNS increased, n=19) and non-progressive (CMTNS not increased, n=11), and the classifier had a specificity of 100% and a sensitivity of 63.2% in the detection of disease progression ([figure 4C](#)). Importantly, five out of the six disease progression biomarkers (*GSTT2*, *CTSA*, *PPARG*, *CDA*, *ENPP1*) are also part of the disease severity biomarker cluster of the 266 patients with CMT1A cohort ([figure 1B,C](#)), further supporting their eligibility as biomarkers in CMT1A.

Thus, skin biopsy is a valuable source for the identification of biomarkers in CMT1A ([figure 2](#), [table 2](#)). Moreover, we identified six genes with changing expression over time that may serve as disease progression biomarkers and could directly be implemented into ongoing clinical trials ([figure 3](#)).

DISCUSSION

Despite promising results from experimental therapies in animal models, CMT1A remains a disease without known cure. Clinical trials are hampered by the slow CMT1A disease progression and the availability of outcome measures with limited sensitivity to

Table 1 Patient characteristics. This table holds descriptive statistics about the main clinical parameters of all the patients in the study cohort. The statistics are computed separately within each contributing clinical center. Grey: basic characteristics; white: CMT scores; green: subscores to the CMTNS; orange: secondary parameters.

Parameter	Level	Czech Republic	France	Italy	Belgium	Spain	UK	Germany/Munich	Germany/Göttingen	p Value
n		56	9	10	26	50	20	40	55	
Gender	Female	37 (66.1%)	5 (55.6%)	6 (60.0%)	17 (65.4%)	33 (66.0%)	15 (75.0%)	15 (37.5%)	27 (49.1%)	0.04
	Male	19 (33.9%)	4 (44.4%)	4 (40.0%)	9 (34.6%)	17 (34.0%)	5 (25.0%)	25 (62.5%)	28 (50.9%)	
Age at Examination (years)	Mean±SD	40±12	46±18	54±16	51±16	39±14	46 ± 12	44±14	43±12	0.01
BMI (kg/(m ²))	Mean±SD	25±5.1	25±4.8	24±3.1	25±4.4	25±4.4	27±5.6	24±4.8	26±5.5	0.32
CMTNS (0/.../36)	Mean±SD	14±4.6	13±4.5	13±7.5	17±4.2	12±4.3	16±4.9	14±5.1	14±5	<0.01
CMTNS_full (0/.../60)	Mean±SD	27±8.4	22±8.7	25±11	28±8.4	19±6.7	36±11	24±8.4	26±6.6	<0.01
CMTNS_sig (0/.../48)	Mean±SD	22±6.2	20±6.3	20 ± 9.4	24±6.4	16±5.5	25±6.2	20±6.7	22±5	<0.01
CMTNS_mod 0/.../32]	Mean±SD	14±4.6	12±5.1	12 ± 5.6	16±5.4	8.1±4.2	16±4.5	13±5.1	16±3.8	<0.01
Sensory Symptoms (0/1/2/3/4)	Mean±SD	1.1±1.1	0.78±0.83	0.56±0.73	1.3±1	0.58±1.1	2.1±0.64	1.3±0.99	1.7±0.83	<0.01
Motor Symptoms Legs (0/1/2/3/4)	Mean±SD	1.2±0.9	1.1±0.78	1.4±1.1	2±1.2	0.84±0.68	1.6±0.82	1.6±1.1	1.3±0.96	<0.01
Motor Symptoms Arms (0/1/2/3/4)	Mean±SD	0.57±0.57	0.67±0.71	0.56±0.73	1.1±0.56	0.38±0.64	1.1±0.51	0.62±0.71	0.81±0.74	<0.01
Pin Sensibility (0/1/2/3/4)	Mean±SD	1.4 ± 0.85	1.7±1	1.1±1.3	1.7±0.88	0.74±1.1	1.6±0.81	1.1±0.81	1.4±0.91	<0.01
Vibration (0/1/2/3/4)	Mean±SD	2.4±0.88	1.3±0.5	1.8±1.4	2.8±0.51	3.7±1	1.4±0.69	1.7±0.86	2±1.2	<0.01
Strength of Legs (0/1/2/3/4)	Mean±SD	1.4±0.87	1.3±1	1.1±1.3	1.4±1	1.4±1.1	1.9±1	1.7±1.2	1.4±0.99	0.32
Strength of Arms (0/1/2/3/4)	Mean±SD	0.79±0.71	1±0.87	1.2±1.3	0.81±0.69	0.2±0.4	1.3±0.86	0.67±0.87	0.48±0.78	<0.01
Ulnar CMAP (Median) (0/1/2/3/4)	Mean±SD	1.7±1.1	1.4±1.3	1.7±1.3	2±0.87	1.2±1.1	1.1±1.2	2.2±1	2.1±1.1	<0.01
Ulnar SNAP (Median) (0/1/2/3/4)	Mean±SD	3.4±0.99	3.3±0.5	3.8±0.67	3.6±0.94	3.3±0.85	3.9±0.45	3.5±0.55	2.9±1.1	<0.01
10m Walk Test (sec)	Mean±SD	7.8±1.7	8±2.8	13±6.4	16±18	6.7±1.4	8.7±2.4	6.1±1.7	9.5±3.4	<0.01
9 Peg Hole Test (dominant hand) (sec)	Mean±SD	27±11	27±7.1	25±8.8	33±23	22±4.4	32±14	24±8.9	29±14	<0.01
Fist Grip (dominant hand) (kg)	Mean±SD	128±80	197±121	105±41	147±79	169±77	104±107	196±77	139±75	<0.01
Three Point Grip (dominant hand) (kg)	Mean±SD	54±24	59±31	50±27	48±28	52±24	29±16	55±25	55±26	0.68
Foot Dorsal (dominant foot) (kg)	Mean±SD	51±35	105±78	50±29	51±38	201±75	30±17	36±28	33±21	<0.01
Foot Plantar (dominant foot) (kg)	Mean±SD	75±46	244±118	75 ± 37	11	263 ± 80	56 ± 29	45±25	71 ±46	<0.01

CMAP: compound muscle action potential; SNAP: sensory nerve action potential.

Table 2 Regression coefficients for clinical parameters modeling biomarker expression. Using multiple linear modeling, the effect of each of the CMT scores as well as each primary and secondary parameter (rows) on the expression of each of the potential biomarkers (columns) was assessed while controlling for effects from the co-variables centre, plate, age, gender and BMI. The table shows the regression coefficients from the individual models. The modeling was done on the Ct scale that is negative logarithmic to the gene expression, thus: regression coefficient < 0: positive effect on gene expression, regression coefficient > 0: negative effect on gene expression. Significant (on a non-adjusted 5% level) values are printed in bold.

	SPRR1A	PPARG	NRG1-I	MUCL1	GSTT2	GSTT1	GSTA4	GRIA1	FN3KRP	FN1	ENPP1	CTSA	CRISP3	CDA	BGN	ANPEP
CMTNS	0.05	0.04	0.03	0.07	0.03	0.05	0.03	0.04	0.04	0.03	0.04	0.03	0.07	0.04	0.04	0.03
CMTNS_full	0.03	0.03	0.01	0.03	0	0.04	0.02	0.01	0.01	0.01	0.02	0.01	0.04	0.02	0.02	0.01
CMTNS_sig	0.04	0.04	0.02	0.05	0.02	0.08	0.03	0.03	0.03	0.02	0.03	0.03	0.05	0.03	0.03	0.02
CMTNS_mod	0.06	0.05	0.02	0.04	0.02	0.1	0.03	0.04	0.04	0.03	0.03	0.04	0.05	0.04	0.04	0.03
Sensory Symptoms	0.01	0.09	0.17	0.13	-0.01	0.3	0.04	0.08	0.01	0.11	0.11	0.04	0.11	0.04	0.08	0.07
Motor Symptoms Legs	0.22	0.24	0.12	0.16	0.13	0.52	0.13	0.22	0.19	0.08	0.12	0.16	0.17	0.24	0.16	0.1
Motor Symptoms Arms	0.4	0.13	0.02	0.11	0.07	0.3	0.1	0.07	0.14	0.01	0.16	0.05	0.24	0.08	0.13	0.05
Pin Sensibility	0.11	-0.02	0.08	0.06	-0.15	0.25	0.03	-0.01	-0.02	0.08	0.05	-0.11	0.04	-0.02	0.08	-0.05
Vibration	0.07	0.23	0.14	0.34	0.24	0.2	0.09	0.26	0.23	0.1	0.31	0.23	0.36	0.26	0.13	0.15
Strength of Legs	0.31	0.3	0.13	0.14	0.22	0.25	0.15	0.28	0.23	0.16	0.23	0.22	0.31	0.3	0.26	0.21
Strength of Arms	0.27	0.12	0.1	0.14	0.07	0.42	0.2	0.04	0.04	0.09	0.1	0	0.34	0.08	0.13	0.01
Ulner CMAP (Median)	0.07	0.1	-0.03	0.12	0.13	-0.04	0.11	0.08	0.12	0.08	-0.04	0.07	0.08	0.05	0.02	0.05
Ulner SNAP (Median)	0.01	-0.03	0.14	0.64	0.08	0.13	0.06	-0.07	0.02	0.11	0.02	-0.01	0.39	0.03	0.14	0.07
10m Walk Test	0.15	0.09	-0.04	-0.06	-0.05	0.2	0.03	0.08	0.1	0.01	0.07	0.06	0.06	0.07	0.03	0.03
9 Peg Hole Test (dominant hand)	0.12	0.04	0.08	0.21	0.03	0.32	0.07	0.01	0.05	0.11	0.05	0.06	0.16	0.05	0.13	0.08
Fist Grip (dominant hand)	0.15	0.12	0	0.15	-0.01	0.07	0.06	0.07	-0.01	0.1	0.14	0.05	0.21	0.07	0.12	0.07
Three Point Grip (dominant hand)	0.02	0.06	-0.06	0.15	-0.05	0.08	0.05	0.04	0	0.08	0.07	-0.02	0.14	0	0.03	0.03
Foot Dorsal (dominant foot)	0.12	0.18	0.04	0.27	0.11	-0.05	0.09	0.16	0.12	0.14	0.15	0.12	0.33	0.15	0.17	0.17
Foot Plantar (dominant foot)	0.19	-0.04	0.01	-0.11	-0.1	0.09	0.1	-0.11	-0.01	-0.02	-0.2	-0.09	0.12	0.07	-0.05	-0.08

BMI, body mass index, CMT, Charcot-Marie-Tooth.

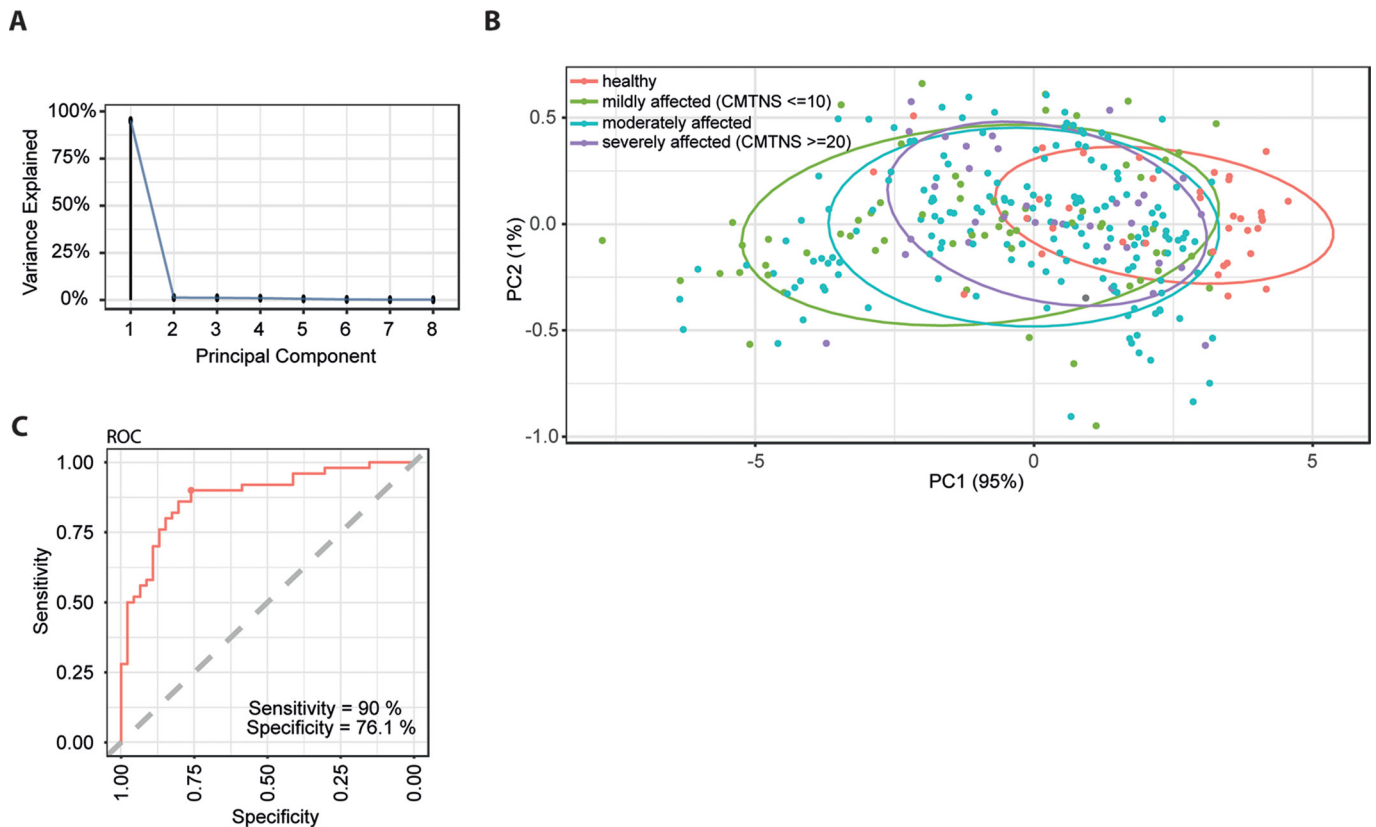


Figure 2 Cutaneous biomarker expression separates Charcot-Marie-Tooth disease type 1A (CMT1A) severity and controls. (A) Principal component analysis (PCA) was performed on the PCR data of the biomarker cluster identified in figure 1B,C. A scree plot shows the percentage of the explained variance for each of the principal components (PCs). PC1 captures most of the information and explains 95% of the observed variance while PC2 explains already only 1% of the observed variance. (B) Mapping of the samples on the first two PCs from (A) with PC1 on the x axis and PC2 on the y axis. All samples are colour coded by their CMT disease severity status (red: healthy, green: mildly affected (Charcot-Marie-Tooth neuropathy score (CMTNS) ≤ 10), blue: moderately affected ($10 > \text{CMTNS} < 20$), purple: severely affected (CMTNS ≥ 20). The ellipses are 70% probability ellipses assuming two-dimensional normally distributed data drawn separately for each of the four groups. (C) A support vector machine (SVM) classifier separating mild and severe CMT cases based solely on gene expression profiles of the eight biomarkers (figure 1B,C) was trained. Shown are the results from a 10 times repeated 10-fold cross-validation. The presented receiver operating characteristic curve plots sensitivity (y axis) versus specificity (x axis). A perfect classifier would reach the top-left corner (100% sensitivity and 100% specificity). The closest point (Youden index) of the trained SVM reaches 90% sensitivity and 76.1% specificity.

change.^{30 37 38} Next to improvement of clinical outcome measures, molecular biomarkers would add powerful tools to assess and track disease progression in clinical trials. Molecular biomarkers may ideally derive from body fluids or minimally invasively sampled tissue. Although it is not known so far whether CMT1A is reflected by alteration of body fluid characteristics which then could be harnessed for biomarker development, skin biopsies have proven powerful tools to study CMT1A.^{39–41} In a preliminary study, we could identify CMT rat-derived cutaneous disease severity biomarkers in skin biopsies from a small cohort of 46 patients with CMT1A. By applying a post hoc multiple linear regression model, in this study we were able to show that 47% of the observed variance in disease severity (as measured by the CMTNS) can be predicted by gene expression and age.¹⁶ The development of applicable biomarkers of disease severity in CMT1A, however, requires a four-step fit-for-purpose approach (see section ‘Introduction’).^{32 33} To validate the basic assay performance (fit-for-purpose step 2), we initiated a worldwide prospective study and nine centres in Europe and the USA contributing clinical data and skin biopsies from 266 patients with CMT1A. The expression of 16 previously identified potential cutaneous disease severity biomarkers was tested in skin biopsies from the new patient cohort and we confirmed that disease severity is indeed reflected by cutaneous gene expression (figure 2). The number of individual genes, however, that significantly

correlate with the cumulative CMTNS variants (maximum 7/16 genes) is relatively low compared with correlations of biomarker expression with selected single parameters (eg, with ‘strength of legs reached 13/16 hits, table 2). The notion that some items of the outcome measures do in fact reflect the actual disease severity more sensitively than others has been reported previously²⁴ and could explain that biomarker correlations vary between CMTNS variants and subscores. However, cutaneous expression of a cluster of eight genes (*CDA*, *CTSA*, *GRIA1*, *ENPP1*, *ANPEP*, *FN3KRP*, *GSTT2* and *PPARG*) correlates with disease severity (as assessed by the CMTNS) with very high sensitivity (90%, figure 2C) confirming that expression analyses are suitable to develop biomarkers (fit-for-purpose step 2). A major caveat, however, is the difficulty to improve outcome sensitivity solely on the basis of a rather insensitive anchor as the CMTNS. To overcome this problem, we performed an analysis of change in gene expression over time (fit-for-purpose step 3). From 16 tested genes, we found *PPARG*, *GSTT2*, *CTSA*, *CDA*, *ENPP1* and *NRG1-I* to show a statistically significant increase in expression over time, all of which (except *NRG1-I*) were identified as a component of a biomarker cluster within the 266 patients with CMT1A cohort (figures 1B,C and figure 4). Independent from the cluster analysis (figure 1), the expression of individual genes showed significant correlations with various CMTNS-related items (table 2). For example, the progression biomarker

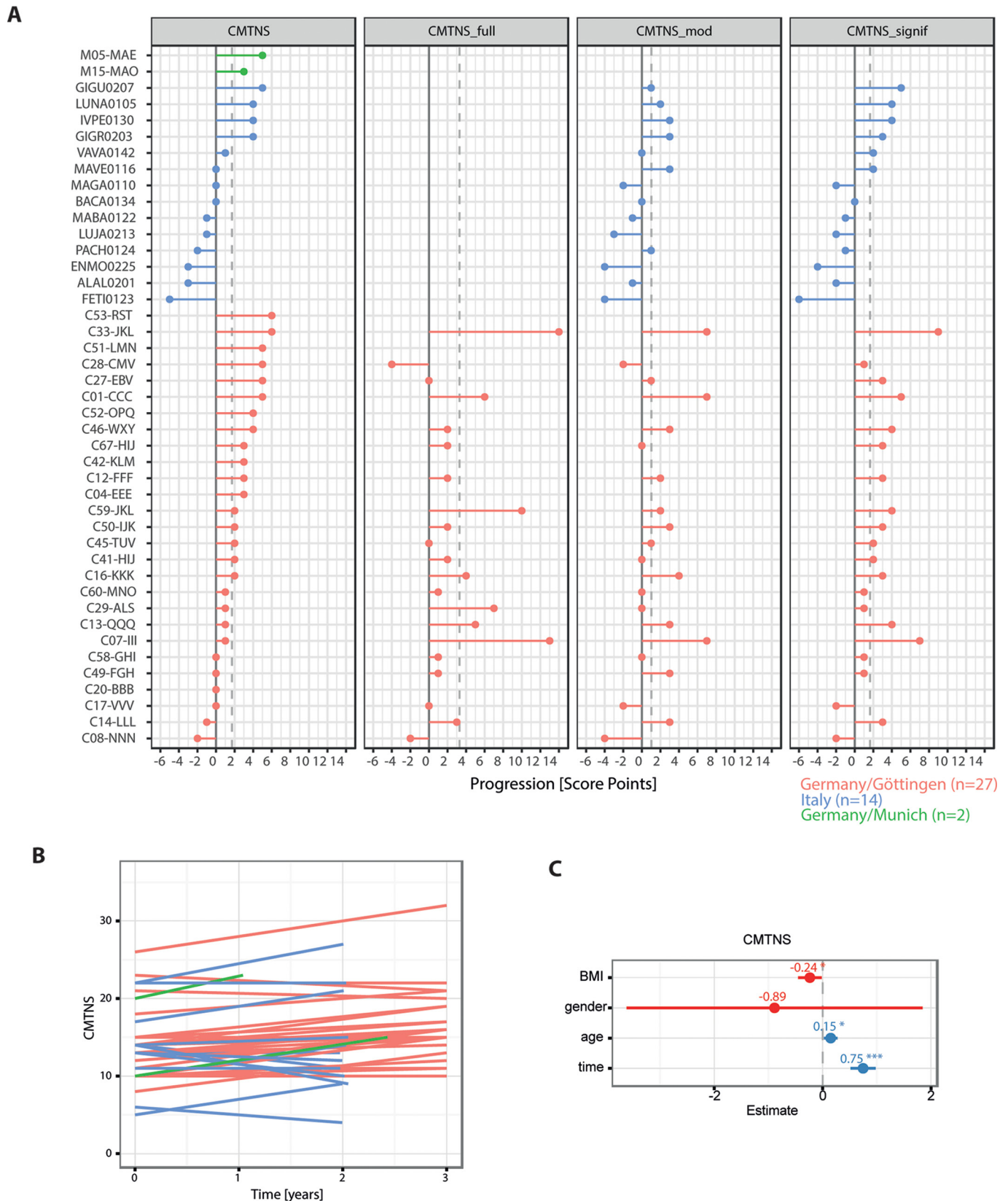


Figure 3 Disease progression is clinically detectable over a 2–3-year time period. (A) Progression of Charcot-Marie-Tooth (CMT) disease measured by the scores CMTNSv1, CMTNS_full, CMTNS_mod and CMTNS_signif with patients (coloured by contributing centre) on the y axis and the score difference on the x axis where positive difference correspond to higher scores at the second examination. (B) The change of Charcot-Marie-Tooth neuropathy score (CMTNS) is shown over time. The time between measurements ranges from 2 to 3 years (with one exception of only 1-year time difference). Each line represents one patient. (C) Regression coefficients with 95% CIs from the fit of a linear mixed effect model for CMTNS with fixed effects. body mass index (BMI), gender, age, and time and a random effect accounting for the repeated measures in the patients. The factor ‘time’ has a significant influence on the CMTNS with an estimated increase in CMTNS of 0.75 per year.

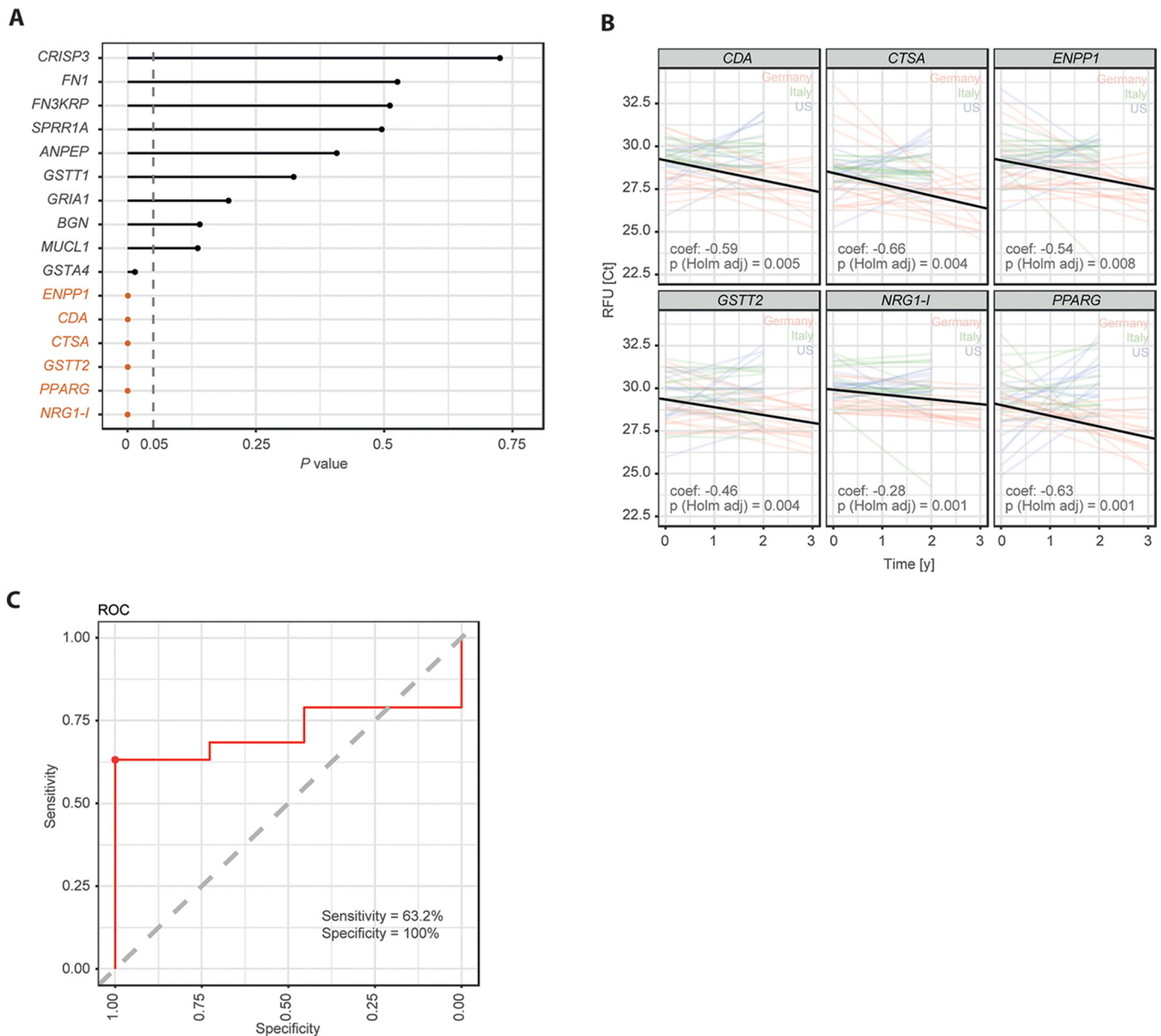


Figure 4 Cutaneous expression of selected biomarkers changes with disease progression. (A) Mixed effects repeated measures regression models were fit per gene to the expression data. The effect of time was assessed controlling for body mass index, gender, age and centre effects. Shown are the p values for the time effect (orange: adjusted p value significant) for all genes. (B) Shown is the progression data underlying the regression models from (A) displayed as Ct value (gene expression) as a function of time. Each coloured line represents one patient and the black lines show the average patient at time 0 progressing by the modelled time effect. (C) A predictive random forest model that classifies patients into progressive patients (Charcot-Marie-Tooth neuropathy score (CMTNS) increases) and non-progressive patients (CMTNS does not increase) was trained on the six genes with significant change in expression over time. Data were available for 19 progressive and 11 non-progressive patients. The consensus receiver operating characteristic (ROC) curve from a 10 times repeated 10-fold cross-validation reveals an area under the curve of 0.74 and the sensitivity/specificity at the Youden index (marked with a dot) are estimated to be 63.2%/100%.

NRG1-I (figure 4A,B) correlates with the CMTNS was the only gene found to correlate with sensory symptoms (table 2), although this gene was not part of the identified cluster. However, on the basis of the insensitive CMTNS, the potential disease progression biomarkers display a sensitivity of 63.2% (figure 4C). Here, the only moderate sensitivity of progression biomarkers towards clinical impairment may directly result from the insensitive CMTNS anchor. The maximum specificity of 100%, in turn, demonstrates that non-progressive patients were easily detectable by biomarker expression in the examined patient cohort. As a consequence, patients which respond to a therapy (and thereby display a halt in disease progression) could be identified more easily with the help

of the here reported disease progression biomarkers. The validation of these biomarkers in the disease progression of further patients with CMT1A in natural history studies (also by next-generation sequencing approaches) may show whether a higher sensitivity can be achieved and may reveal further evidence for the potential practical use of these biomarkers.

Enrichment analyses of the eight clustered genes (table 3) revealed an involvement in metabolism (Reactome Pathway), negative regulation of cell growth (Gene Ontology, biological process), vacuolar/lysosomal membrane (Gene Ontology, cellular component) and exopeptidase activity (Gene Ontology, molecular function). This functional annotation may fit to the

Table 3 Gene enrichment analysis and functional annotation of CDA, CTSA, GRIA1, ENPP1, ANPEP, FN3KRP, GSTT2 and PPARG. Enriched terms are shown for the databases a) Reactome Pathway, b) Gene Ontology (GO), biological process, c) GO, cellular component and d) GO, molecular function. The maximum 20 significant hits are displayed with term, database identifier, gene set overlap, adjusted p-value, Z-score, combined score and genes involved (only hits with more than one gene)

Term	Identifier	Overlap	Adjusted p-value	Z-score	Combined Score	Genes
Reactome pathway						
Metabolism	R-HSA-1430728	5/1908	0.01167	-2.25	9.99	CTSA;CDA;ENPP1;GSTT2;PPARG
Metabolism of proteins	R-HSA-392499	4/1074	0.01167	-2.14	9.54	CTSA;GRIA1;ANPEP;FN3KRP
Post-translational protein modification	R-HSA-597592	3/521	0.01428	-2.18	9.25	CTSA;GRIA1;FN3KRP
Asparagine N-linked glycosylation	R-HSA-446203	2/259	0.03554	-2.08	6.95	CTSA;GRIA1
GO biological process						
Negative regulation of cell growth	GO:0030308	3/150	0.00619	-2.25	11.44	CDA;ENPP1;PPARG
Negative regulation of growth	GO:0045926	3/221	0.00984	-2.31	10.68	CDA;ENPP1;PPARG
Regulation of cell growth	GO:0001558	3/322	0.01999	-2.34	9.17	CDA;ENPP1;PPARG
Receptor metabolic process	GO:0043112	2/71	0.02359	-2.12	7.94	GRIA1;PPARG
Regulation of fat cell differentiation	GO:0045598	2/87	0.02832	-2.32	8.26	ENPP1;PPARG
Cellular response to peptide hormone stimulus	GO:0071375	2/261	0.03832	-2.39	7.80	ENPP1;PPARG
Cellular response to peptide	GO:1901653	2/273	0.03832	-2.38	7.76	ENPP1;PPARG
Cellular response to insulin stimulus	GO:0032869	2/195	0.03832	-2.36	7.68	ENPP1;PPARG
Response to insulin	GO:0032868	2/246	0.03832	-2.33	7.60	ENPP1;PPARG
Protein maturation	GO:0051604	2/222	0.03832	-2.29	7.46	ANPEP;PPARG
Receptor-mediated endocytosis	GO:0006898	2/152	0.03832	-2.27	7.41	GRIA1;ENPP1
Protein processing	GO:0016485	2/201	0.03832	-2.24	7.32	ANPEP;PPARG
Endocytosis	GO:0006897	2/323	0.03832	-2.24	7.30	GRIA1;ENPP1
Cellular response to organonitrogen compound	GO:0071417	2/411	0.03955	-2.40	7.75	ENPP1;PPARG
Cellular response to nitrogen compound	GO:1901699	2/438	0.03955	-2.37	7.64	ENPP1;PPARG
Response to peptide	GO:1901652	2/384	0.03955	-2.31	7.47	ENPP1;PPARG
Response to peptide hormone	GO:0043434	2/364	0.03955	-2.31	7.47	ENPP1;PPARG
Cellular response to hormone stimulus	GO:0032870	2/462	0.03955	-2.29	7.39	ENPP1;PPARG
Negative regulation of phosphate metabolic process	GO:0045936	2/458	0.03955	-2.26	7.30	CDA;ENPP1
Anion transport	GO:0006820	2/443	0.03955	-2.14	6.91	ENPP1;PPARG
GO cellular component						
Vacuolar part	GO:0044437	3/285	0.00730	-2.32	11.42	CTSA;ANPEP;ENPP1
Lysosomal membrane	GO:0005765	2/196	0.04628	-2.26	6.93	ANPEP;ENPP1
Vacuolar membrane	GO:0005774	2/208	0.04628	-2.25	6.93	ANPEP;ENPP1
GO molecular function						
Exopeptidase activity	GO:0008238	2/111	0.03047	-2.49	8.69	CTSA;ANPEP
Zinc ion binding	GO:0008270	4/1256	0.03047	-2.31	8.05	CDA;ANPEP;ENPP1;PPARG

reported impact of the duplicated disease gene PMP22 on lysosomal alteration on overexpression in CMT1A.⁴²

Taken together, the validity of the identified biomarkers was tested twice, first as disease severity biomarkers in a cohort including 266 patients with CMT1A and 37 healthy controls and, second, as progression markers in an additional longitudinal cohort including 45 patients with CMT1A. We confirmed that gene expression analysis of the skin is suitable for the assessment of the disease severity, and that the cutaneous expression of individual genes is changing over time in patients with CMT1A. We suggest that the expression of five genes that were part of an identified biomarker cluster and correlated with disease progression over time (*CDA*, *CTSA*, *ENPP1*, *GSTT2*, *PPARG*) is a valid set of biomarkers in CMT1A. When implemented and successfully validated in ongoing clinical trials, the here reported biomarkers could markedly accelerate the development of a therapy for CMT1A.

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Contributors RF monitored data collection, supervised sample processing and qPCR analyses, drafted the statistical analysis plan, and wrote and revised the paper. He is guarantor. AL collected and curated all data, drafted the statistical analysis plan, analysed the data, and wrote and revised the paper. He is guarantor. MM implemented the biomarker trial and collected clinical data in Göttingen, Germany, and monitored collection of all clinical data. CE performed qPCR analyses. TB supervised statistical analyses. MM, AS, ALPN, JB, BSW, TJS, TP, NGA, DC, JH, RM,

WP, MCW, CMT-TRIAAL, JYH, OD, AS, JB, PDJ, MES, RH, DP, PS, PY and MWS were involved in patient recruitment, patient assessment and skin biopsy sampling. MM, TJS, TP, DC, BSW, NGA, MCW, WP, PY and MWS implemented the study in Germany. AS, AS, DP and CMT-TRIAAL implemented the study in Italy. ALPN and JB implemented the study in Spain. JH, RM and PS implemented the study in the Czech Republic. JYH and OD implemented the study in France. JB and PDJ implemented the study in Belgium. RH implemented the study in the UK. MES implemented the study in the USA. MWS initiated, coordinated and supervised the biomarker study. He is guarantor.

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Correction: Biomarkers predict outcome in Charcot-Marie-Tooth disease 1A

Fledrich R, Mannil M, Leha A, *et al.* Biomarkers predict outcome in Charcot-Marie-Tooth disease 1A. *J Neurol Neurosurg Psychiatry* 2017;88:941–52. doi: 10.1136/jnnp-2017-315721.

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