1   Identification of novel plasma proteomic biomarkers of Dupuytren Disease

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Abstract

Dupuytren Disease (DD) is a chronic progressive disease that can cause disabling hand

deformities. The most common treatments have high rates of complications and early

recurrence. Dupuytren lacks a staging biomarker profile to develop preventive therapeutics to

improve long-term outcomes. This multi-omic study aimed to create a DD blood proteomic

biomarker profile by comparing DD plasma to a healthy control group. We measured circulating

collagen metabolism peptides and found normal Collagen I synthesis but impaired Collagen I

degradation in DD. We measured 6995 serum protein aptamers and identified 68 proteins with

statistically significant differences from the control group. We developed two Diagnostic

Proteomic Risk Scores (DPRS) based on hypothesis-free and hypothesis-based analyses. In

independent data, our hypothesis-free and hypothesis-based DPRS distinguished Dupuytren

from control subjects with 76.5% and 70.6% accuracy, respectively. Our hypothesis-based

DPRS also distinguished DD subjects with different disease progression rates based on subject

age at the time of their first corrective procedure (p=0.0018). This pilot study is the first to

provide evidence that Collagen I accumulation in DD is due to impaired degradation rather than

increased collagen synthesis. It also describes novel DPRS that have potential use as

diagnostic and staging biomarker panels for Dupuytren disease.

38   Introduction

Dupuytren disease (DD) is a progressive fibrotic condition that often results in finger deformities

called Dupuytren contracture (DC). DD affects an estimated 10 million Americans. DC is

associated with loss of productivity, higher healthcare costs, and a greater prevalence of benign

diseases[1].

The greatest need in DD care is preventing DC progression and recurrence. As with other

fibrotic diseases, developing preventive treatments requires a biomarker profile for staging and

measuring therapeutic response. The authors have not found published reports with statistically

significant evidence of preventing DC recurrence or a biomarker tool to develop such

treatments. This pilot study aimed to develop preliminary data on a diagnostic blood biomarker

profile of DD.

DC is a common outcome of DD. Of those diagnosed with DD at age 65, one in four men and

one in eight women will require at least one corrective procedure over their lifetime[2]. Current

treatments are palliative, often with partial or temporary improvement[3]. Surgery for DC has a

greater risk of major complications than other common elective hand operations[4-7], and

recurrent DC is the most common diagnosis leading to elective finger amputation[8].

Genetic factors account for about 80% of the risk of developing DD[9]. Family history, age, and

gender have strong associations. A genetic risk score (GRS) based on 11 SNPs correlates with

DD diagnosis[10], and a weighted GRS based on 26 SNPs correlates with post-procedure

recurrence[11].

DD affects the palm skin, subcutaneous tissues, and fascia[12]. Early disease is cellular, with

fibroblasts, myofibroblasts, and perivascular clusters of myofibroblasts, pericytes, and

mononuclear immune cells[13]. Early changes include local variations of immune[13],

hematopoietic, and mesenchymal cell markers[12], cytokines, signaling factors[14], and

extracellular matrix components[15, 16]. Progression transforms tissues into nearly acellular,

densely packed collagen.

Physical changes are the basis for DD diagnosis, staging, and treatment, but physical findings

don't predict progression[17]. There are no standard noninvasive tests of DD biological activity.

Biopsy is problematic because it can trigger disease activity and progression. In addition, DD-

related tissue findings do not always correlate with blood measurements[18-20], and DD tissue

gene expression does not correlate with local protein levels[21]. Blood remains the most

promising source for a disease activity profile.

We conducted a discovery study to compare levels of circulating proteins and peptides in

subjects with DD versus controls using ELISA, aptamer, and mass spectrometry methods. We

found statistically significant differences in disease versus control cohorts in independent

analyses of peptides and proteins. We also found statistically significant differences in protein

aptamers in DD subjects with different rates of clinical progression of untreated disease.

Materials and methods

Sex as a biological variable

77 Our study involved male and female subjects. Sex was not considered a biological variable.

Subjects

79 We selected disease cohort subjects from the International Dupuytren Database (IDDB), a

80 patient-reported registry developed by the Dupuytren Research Group (DRG) in collaboration

81 with the National Databank of Rheumatic Diseases (NDB). Control subjects were enrolled by

82   word of mouth and from the NDB database (https://dupuytrens.org/enroll-in-the- iddb-2/). All

83   subjects in the IDDB met the entry criteria and signed an Informed Consent Form.

Entry criteria

The DD cohort met criteria based on clinical evidence of DD with strong traditional predictors of

recurrent deformity after a corrective procedure: age of disease onset younger than 50; two or

more first-degree relatives with DD; current DD nodules; three or more separate locations of DD

or two palm locations and Ledderhose disease (a related disorder). Control cohort subjects

were healthy individuals with no clinical evidence of DD, no personal history of DD or

Ledderhose disease, and no family history of DD.

We confirmed study eligibility by requiring candidates to upload recent images of each hand's

palmar and dorsal sides. The Principal Investigator/ senior author confirmed entry criteria and

reviewed images to confirm clinical evidence of active DD in the disease cohort and the lack of

these findings in the control cohort.

Blood Sample Collection

The study used a mobile phlebotomy service to collect blood specimens (Phlebotek Solutions

Corporation, Fort Lauderdale, FL). Once confirmed that the subject met all study entry criteria,

they were contacted by DRG and later by Phlebotek to schedule their blood draw. The

phlebotomist visited the subjects in their homes and collected two tubes of blood. The last 16

blood draws included a third 2.5ml PAXgene tube collection for additional genomic testing.

Processing of the EDTA tubes for proteomic testing was standardized. Immediately following

collection, blood samples were centrifuged for 15 minutes at 2000 g to generate plasma. The

plasma was harvested from the top of the container and pooled before aliquoting into storage

cryotubes. The last 500 µL of plasma was discarded to avoid contamination with platelets.

Samples were immediately packaged for shipping on dry ice.

Number of subjects

One hundred subjects were planned (50 DD/50 Control). Blood collection launched just as the

COVID pandemic began (i.e., February 2020). Due to COVID-related issues, we amended the

protocol to enroll 50 subjects. Forty-seven subjects were enrolled;46 subjects had samples

drawn. One sample was disqualified due to failed shipping refrigeration. Samples from the

remaining 45 subjects (25 DD, 20 Control) had mass-spectrometry (Bioproximity, Manassas,

VA). Twenty-seven subjects (13 DD, 14 control) had second blood samples drawn six to 14

months after their first draw for mass-spectrometry testing by a second lab (MacCoss Lab,

Seattle, WA). Additional proteomic testing was conducted on remaining aliquots: 1) n= 27

redrawn samples had ELISA tests for collagen metabolites (Nordic Bioscience, Herlev, DK); 2)

n= 72 from both initial and redraw specimens had Aptamer-based testing (SomaLogic, Boulder,

CO). Supplementary Fig. S1 outlines the distribution of samples and laboratories.

Blinding

119   All subjects were assigned and identified by participant numbers. Investigators were blinded to

120   the sample group allocation. All samples remained blinded throughout the processing and data

121   collection in the laboratories.

Assays

Nordic Bioscience collagen markers

Circulating collagen synthesis and degradation markers were measured using proprietary

assays[22]. Synthesis biomarkers included collagens I (PRO-C1), II (PRO-C2), III (PRO-C3), IV

(PRO-C4), V (PRO-C5), VI (PRO-C6), VII (PRO-C7), and IX (PRO-C9). Circulating degradation

markers were measured for collagens I (C1M), II (CALC2, C2M), III (C3M, CTX-III), IV (C4M), VI

(C6M), and VII (C7M). All biomarkers were assessed by competitive ELISAs and validated to

measure in human plasma samples. The inter- and intra-assay coefficients of variation were

>15% and 10%, respectively, for all assays.

Mass spectrometry

A novel untargeted proteomics assay using a large-scale liquid chromatography-tandem mass

spectrometry (LC-MS/MS) approach called data-independent acquisition (DIA) was used to

collect proteomic data. Each sample (10 µl plasma), following tryptic digestion, was diluted

50:50 by the addition of 4% sodium dodecyl sulfate (SDS) buffer to a final concentration of ~2%

SDS. Approximately 200 µg of protein was added to 96-well plates, and the remainder of the

sample preparation was performed using the single-pot, solid-phase enhanced sample

preparation for proteomics methodology (SP3) [23, 24]. The entire process of protein capture,

reduction and alkylation, protein cleanup, digestion, and peptide cleanup was performed using a

KingFisher magnetic bead robot[24]. Peptides were reconstituted in HPLC starting aqueous

buffer (5% acetonitrile and 0.1% formic acid) with analysis performed on the high-resolution

quadrupole-orbitrap hybrid mass spectrometer (Exploris 480, Orbitrap Fusion Lumos, or

Orbitrap Fusion Eclipse).

Data was acquired in a single, 2-hour LC-MS/MS run (90 min HPLC gradient) on a Thermo

Quadrupole-Orbitrap mass spectrometer. Crucially, the DIA method comprehensively samples

all peptides between 400 and 1000 m/z using a repeated cycle of 75 MS/MS scans with an 8

m/z-wide isolation width. Using a multiplexing scheme based on staggering the isolation window

in alternating cycles of MS/MS scans, the precursor selectivity of the resulting data is 4 m/z after

computational demultiplexing[25]. From the resulting data, fragment ion chromatograms were

extracted for any peptide precursor between 400 and 1000 m/z. To determine which peptides

are confidently detected in each acquisition, the software tool EncyclopeDIA[26] was used to

query the data using a reference chromatogram library generated from a prior analysis of the

tissue. Chromatogram-based relative quantitation was then performed for each peptide

detected. All measurements are made relative to a common reference. In the case of plasma, a

common reference is a pooled standard from 400 age-matched samples from subjects who are

both healthy and span the phenotypes being studied, processed in each sample batch. Within

each 96-well plate, one copy of the common reference standard is placed in each row for a total

of 8 copies per plate[27]. Before quantification, it is confirmed that peptides are stable during the

time scale of the data collection, have a suitable linear response, and have suitable intra- and

inter-data precision.

SomaScan

SomaLogic, Inc. (Boulder, CO, USA) performed proteomic analysis of the plasma samples using

a SomaScan platform that quantifies the presence of 6995 human proteins (secreted proteins,

extracellular domains, and intracellular proteins). These proteins are found in diverse biological

groups such as receptors, kinases, cytokines, proteolytic enzymes and inhibitors, growth

factors, protease inhibitors, hormones, and structural proteins[28]. Most of these proteins

involve pathways relating to signal transduction, stress response, immune processes,

phosphorylation, proteolysis, cell adhesion, cell differentiation, and intracellular transport[29].

Proteins were measured using the Slow Off-Rate Modified Aptamer (SOMAmer)-based capture

array that uses chemically modified single-stranded DNA sequences capable of uniquely

recognizing individual proteins via one or more binding sites. SOMAmers transform a protein

signal to a nucleotide signal that is quantified using relative fluorescence on microarrays. Each

SOMAmer array underwent validation for specificity, upper and lower limits of detection, and

both intra- and interassay variability. Plasma dilutions (0.005%, 0.05%, and 20%) were applied

to capture low-, medium-, and high-abundance proteins. Positive and negative controls were

also positioned on the array to confirm if the experimental procedure was performed correctly.

After the hybridization step, the microarrays were washed and scanned using a laser scanner

that excites the fluorescence of the fluorochrome used in the labeling step. The amount of the

emitted signal is directly proportional to the amount of dye on the microarray. The scanner

measures this quantity and creates a digital image map of the position of each signal based on

the location of its origin. Image analysis software generates a text file from this map describing

the pixel intensity of the spot and the background. This information was processed, and a final

value was generated that summarized the level of each detectable protein on the microarray.

Quality control analysis of the sample and SOMAmer levels involved SOMAmer control arrays

on the microarray and calibration samples. Hybridization controls measured the sample-by-

sample variation in hybridization, while the median signal over all SOMAmer arrays measured

the technical variability. The SomaScan measures were reported as relative fluorescence units

(RFU) in a summary ADAT file.

189   Analysis and statistics

190   Nordic collagen marker analysis

191   Nordic collagen data turnover rates were calculated by performing ratios of the propeptide over

192   the mature peptide (e.g., PRO-C1/ C1M). This ratio is an estimate of the formation/degradation

193   of each collagen type and can be interpreted as a net gain of collagen/loss of collagen

194   remodeling[22]. All comparisons were done using Student's T-test.

195   SomaScan analysis

A series of quality control and data cleaning steps was performed on the SomaScan proteomics

data. All non-human proteins were excluded from the analyses. Samples that did not pass the

default normalization acceptance criteria for all row scale factors were removed. All aptamers

that did not pass the default quality control criteria across all plates/sets were filtered out. The

data had no missing values, and the protein values were log-transformed. No further

normalization was performed.

We identified differentially expressed aptamers using linear regression of each log-transformed

protein as the outcome and disease status as the main predictor. We adjusted each analysis by

age and sex. We analyzed all repeated measurements of each individual and employed

generalized estimating equations to account for technical and biological variability using the gee

package in R. Due to the high number of aptamers tested in the SomaScan assay (6995), we

were at a statistical testing disadvantage regarding multiple corrections. Therefore, we

calculated adjusted p-values using Benjamini-Hochberg methods. Our SomaScan analyses only

report proteins with statistically significant differences from control at an FDR-adjusted p-value <

0.25.

Mass spectrometry analysis

212   For the MS protein data, all zeroes were replaced with NAs across samples, and protein values

213   were log-transformed. The MS dataset was subset based on proteins that overlapped with the

214   signature proteins associated with DD previously identified using the SomaScan proteomics

215   data. We used a generalized estimation equation model to account for the repeated blood

216   draws, adjusted by age and gender.

Proteomic Risk Score calculations

To prepare the data for the Diagnostic Proteomic Risk Score (DPRS) calculation, we first

removed all highly correlated proteins (Pearson correlation > 0.8). 3369 out of 6995 aptamers

were highly correlated and were excluded from the DPRS calculation. The DPRS calculation

contained the following two steps. First, we fit a logistic regression for each protein with disease

status (i.e., case or control) as the outcome variable adjusted by age and gender. We used a

generalized estimation equation model for repeated measurements of two independent blood

draws. We included the same subject's repeated measurements and used the generalized

estimation equations model with the gee R package[30] to estimate the regression coefficients.

Next, the DPRS was computed using summary statistics from each regression model. Only the

proteins with statistically significant differences from control at an FDR-adjusted p-value < 0.25

were included in the score calculation. The DPRS was calculated by summing the beta

estimated effect for all significant proteins, each multiplied by normalized protein abundance

and divided by its standard error. Finally, we fit a logistic regression to examine the association

of the computed DPRS with DD onset. We split the dataset into training (75%) and testing sets

utilizing the caret R package for a leave-one-out cross-validation approach for model fitting and

assessment (25%) sets[31]. We used R version 4.2.1 for all analyses.

Pathway analysis

STRING (12.0) is a web tool that provides and illustrates protein-protein interactions and

networks[32]. STRING integrates different types of evidence from public databases, including

genomic context, high-throughput, and biochemical interaction experiments, known protein

complexes, computational predictions, and text-mining from scientific literature. Given a list of

gene names, STRING constructs protein-protein interaction networks where each gene is

considered as a node, with edges representing the predicted functional associations.

Networks were generated using the whole genome as the statistical background, and interaction

sources included text mining, experiments, databases, co-expression, neighborhood, gene

fusions, and co-occurrence. Interaction scores were created by combining the probabilities from

all evidence channels and correcting for the probability of randomly observing an interaction. All

nodes and interactions were computed with a required medium confidence interaction score

(0.400) and an FDR<=0.05.

Study Approval

The Institutional Review Board Ascension Via Christi Hospitals, Wichita, Inc., approved this

study and the associated informed consent form. The written informed consent was signed and

received before subject participation. This written informed consent included a description of the

use of the photographs/ images to be reviewed for study entry. The records of these signed

informed consents have been retained. All samples remained blinded throughout the processing

and data collection in the laboratories. All methods were performed in accordance with relevant

guidelines, regulations, and the Declaration of Helsinki.

Results

Plasma collagen metabolites in DD

Collagen synthesis and degradation processes release protein fragments specific to the

process, collagen type, and enzyme involved. The ratio of synthesis to degradation-related

fragments correlates with net accumulation or loss of that collagen type. We evaluated plasma

collagen synthesis and degradation markers using the competitive ELISA method (Nordic

Biosciences; Herlev, Denmark) and calculated collagen synthesis-to-degradation ratios[22].

We measured markers of types I, II, III, IV, V, VI, VII, and IX collagen metabolism in blood from a

cohort of healthy controls (n = 14) and subjects with DD (n = 13). We found significantly low

levels of markers of Collagen I degradation C1M (Figure 1A, p = 0.004), Collagen V synthesis

Pro-C5 (Figure 1B, p = 0.046), and Collagen VII synthesis Pro-C7 (Figure 1C, p = 0.048) in the

DD cohort compared to the control group. The remaining collagen biomarkers did not show

significant differences between the disease and control groups (Table 1).

We calculated collagen synthesis-to-degradation ratios for collagen types I, II, III, IV, and V. The

Collagen I synthesis to degradation ratio, Pro-C1/C1M, was significantly greater in the DD

cohort compared to the control group (Figure 1D, p = 0.024). The remaining ratios were not

significant (Col II, III, IV, VI) or could not be calculated because the degradation marker was

either unavailable (Col V, IX) or not detectable in both disease and control cohorts (Col VII).

Differentially expressed blood proteins in DD

We performed two orthogonal unbiased proteomic methods, the aptamer-based SomaScan

(SomaLogic; Boulder, CO) and liquid chromatography-tandem mass spectrometry (LC-MS/MS)

(MacCoss Lab, University of Washington, WA, USA).

We used sixty-nine samples for SomaScan testing after quality control (36 DD and 33 controls).

These represent 42 subjects, 27 of whom had second samples obtained six to fourteen months

after the first (Supplementary Fig. S1). The SomaScan panel measures 6995 protein binding

sites applying to 5444 proteins.

Hypothesis-free approach

The 6995 aptamer measurements identified 54 proteins with statistically significant differences

between DD and control groups, with 24 overexpressed and 30 underexpressed after adjusting

for multiple comparisons (Supplementary Table S1). Figure 2 is a heatmap of these 54 protein

expressions.

We performed pathway analysis of these 54 proteins to identify network relationships using

STRING[32]. We found significant protein-protein interactions (PPI) in 29, with multiple enriched

categories (Figure 3C, p-value = 4.55E-04). Supplementary Table S2 summarizes categories of

enriched gene groups, and Supplementary Fig. S2 provides a visual cross-reference of

functional categories and individual genes. The dominant categories were coagulation and

complement cascades, the extracellular matrix, enzyme inhibition, cell membrane structures,

phosphorylation, and signaling pathways.

Using the SomaScan results, we aimed to develop a hypothesis-free indicator of DD diagnosis

with a calculated Diagnostic Proteomic Risk Score (DPRS) to estimate an individual's disease

risk based on plasma protein abundance.

Hypothesis-free DPRS

We developed a hypothesis-free DPRS estimation based on two of the 54 proteins. These

proteins, OSCAR (overexpressed) and SPART (underexpressed), had a low relative correlation

(Pearson's correlation coefficient = -0.19) (Table 2), suggesting minimal interaction. Using this

hypothesis-free DPRS as a diagnostic measure for DD achieved an accuracy of 79% in the

training set (AUC 0.8) and 76.5% in the test set (Figure 3A), indicating a predictive value for

discriminating between disease and control. Figure 3B shows a box plot of control and DD

DPRS. The DPRS calculated for controls and DD significantly differed by the Wilcoxon test (p-

value = 2.5E-06).

Hypothesis-based approach

We also conducted a hypothesis-based approach with 328 relevant proteins identified from a

literature search and available on the SomaScan platform (Supplementary Table S3). Twenty-

three proteins in this subset had statistically significant differences between DD and control, with

nine overexpressed and fourteen underexpressed after adjusting for multiple comparisons

(Supplementary Table S4).

We performed PPI networks and enrichment analysis on these 23 proteins using STRING. We

found significantly enriched PPI in 21 proteins with multiple enriched categories (Figure 4C, p-

value = 1.35e-07). We found more enrichment categories and greater significance in the

hypothesis-based than the hypothesis-free analysis, which may reflect a higher-yield group,

selection bias, or both. Supplementary Table S5 lists enriched gene categories, and

Supplementary Fig. S3 provides a visual cross-reference between functional categories and

individual genes.

Hypothesis-based DPRS

We developed a hypothesis-based DPRS using 11 of 23 DE proteins identified in the

hypothesis-based analysis. Six genes were underexpressed (*SMAD1, SH3BP2, LCN2, USP8,*

CSNK1G2, SERPINH1) and five overexpressed (*CSMD1, POSTN, ACAN, AOC3, DDR2*). Six

proteins were present in hypothesis-based and hypothesis-free DPRS analyses (Table 3). Our

hypothesis-based DPRS achieved a training accuracy of 78.85% (AUC 0.9) and a test accuracy

of 70.59% in distinguishing Dupuytren from control cohorts (Figure 4A). Figure 4B shows that

DPRS distributions of Control and DD groups differ significantly (Wilcoxon test (p-value = 2.2E-

09).

Evaluation of DRPS for prognosis

We explored the potential of each DPRS to profile DD prognosis using the rate of contracture

progression as an index of biological severity. We segregated our DD samples into subjects

younger (11 specimens) or older (16 specimens) than 50 at the time of their first DC procedure.

We excluded subjects younger than 50 from this analysis. A one-way ANOVA analysis unveiled

a statistically significant variance in the means of hypothesis-driven generated DPRS across

groups with different rates of pretreatment contracture progression (Supplementary Fig. S4, p-

value = 0.0018). Conversely, we found no associations in the DPRS generated through the

hypothesis-free approach.

Mass Spectrometry

337 The LC-MS dataset contained 48 samples: 27 DD and 21 controls. This assay identified 561

338 peptides. No peptides had statistically significant differences in abundance across groups (FDR

339 > 0.25).

Mass-Spectrometry-SomaScan concordance

341 To further validate our SomaScan findings, we evaluated the MS results for concordance. Of the

342 54 significant proteins from the SomaScan data analysis, six were present in the Mass

343 Spectrometry findings: AOC3, POSTN, SERPINC1, TF, PCSK9, and KNG1. We ran the same

344 analyses on these six proteins using Mass Spectrometry data. AOC3, POSTN, and SERPINC1

345 had adjusted p-values (FDR) <0.25 in both Mass Spec and SomaScan analyses (Table 4).

Discussion

This study aimed to discover candidate biomarkers that correlate with the diagnosis of DD as

the first step toward a diagnostic DD blood test.

Dynamic biomarker panels are essential to developing preventive therapeutics for chronic

diseases like DD. The relevance of this study is that DD lacks such biomarkers. Blood is the

most promising unexplored source of proteomic biomarkers for trend analysis.

There are no broad surveys of circulating proteins in DD. In addition, blood findings may not

parallel tissue findings because of rapid degradation, intracellular or membrane-bound

locations, post-translational variants, protein binding, and other mechanisms. Because of this,

new and diverse discovery research is necessary to profile DD-associated blood protein

biomarkers.

There are two common biomarker discovery strategies: hypothesis-free and hypothesis-based.

The hypothesis-free approach (unbiased or non-hypothesis-based) screens many markers and

is the standard in systems biology and discovery research. The hypothesis-based strategy

evaluates selected markers based on published information and is the most common model of

DD research. We used both methods in this study.

Plasma collagen metabolites in DD

Collagen accumulation is the hallmark of DD. DD-affected tissues have unique proportions of

collagen types I[33], III[34], IV[16], and V[33]. The most prominent tissue changes in the

progression from preclinical to end-stage disease are increases in total Collagen I and the

proportion of Collagen III[33].

Our hypothesis-based approach compared collagen synthesis and degradation markers in DD

versus control groups. Collagen turnover releases circulating peptides unique to the synthesis

and degradation of each collagen type. In the absence of disease, synthesis and degradation

markers are tightly coupled. Plasma synthesis to degradation marker ratios can indicate net

collagen gain, loss, or homeostasis[22]. We used ELISA to measure plasma levels of sixteen

collagen metabolism markers involving eight collagen types. We summarize our findings in

Table 1.

The DD cohort had normal Collagen I synthesis marker Pro-C1 levels but statistically significant

decreases in Collagen I degradation marker C1M (Figure 1A), resulting in an elevated

synthesis-to-degradation marker ratio of DD compared to controls (Figure 1D). These findings

suggest DD collagen accumulation follows impaired Collagen I degradation rather than excess

collagen synthesis. Both degradation-resistant collagen variants and reduced collagenase

activity may contribute to reduced collagen degradation.

Dupuytren collagen has two features that increase resistance to enzymatic degradation. Typical

Collagen I fibrils are heterotrimers of two Collagen I α1 strands and one Collagen I α2 strand.

DD-affected tissues have an increased proportion of α1 (I) homotrimer Collagen I[35]. DD

collagen also has more crosslinks stabilized by hydroxylation and glycosylation[36]. Both

homotrimer Collagen I[37] and glycosylation-stabilized crosslinks[38] increase resistance to

enzymatic cleavage.

DD-related dysregulation of collagenases and their inhibitors impairs collagen degradation by

several mechanisms. TIMP1 (Tissue Inhibitor of Metalloproteinase 1) inhibits multiple MMPs

and is elevated in the circulation of DD patients[18]. MMP1, MMP8, MMP13, and MMP14 gene

polymorphisms are associated with Dupuytren disease[39, 40], including MMP14 partial loss-of-

function variants[40]. MMP14 is necessary for MMP2 activation, and both enzymes facilitate

myofibroblast contraction[41].

Two clinical findings support a relationship between impaired collagen degradation and DD. The

first is the rapid onset of DD with metalloproteinase inhibitor treatment[42]. The second is

abnormally slow matrix remodeling in DD. Enzymatic cleavage of collagen fibrils is necessary

for matrix shortening[43], and impaired cleavage would slow this process. Slow matrix

remodeling could explain why DD-related tissue shortening progresses more slowly than wound

healing despite similar biology and why DD surgery has an increased risk of delayed wound

healing[44]. The role of impaired collagen degradation in pathogenesis and intervention is

unclear, and simply normalizing collagen degradation might have the undesirable effect of

accelerating tissue contraction.

The hallmark of fibrotic diseases is collagen accumulation. Increased collagen synthesis

contributes to collagen accumulation in pulmonary fibrosis[45], liver fibrosis[46], primary biliary

cirrhosis[47], renal fibrosis[48], and cardiovascular disease[49]. Our finding of normal levels of

circulating Collagen I synthesis biomarkers in DD suggests that the mechanism of collagen

accumulation in DD is not due to the increased collagen synthesis associated with these

diseases. Although often used interchangeably with increased collagen synthesis, neither

collagen accumulation nor upregulated collagen gene expression is de facto evidence of

increased collagen synthesis. Cell culture studies show increased *COL1A1* and *COL1A2* mRNA

expression in DD-derived fibroblasts[21, 50], but these RNA findings are not associated with

increased collagen I protein production[21]. Our finding of normal Collagen I synthesis and

impaired degradation informs future research because DD drug development has focused on

reducing collagen synthesis rather than increasing degradation[51].

Collagen II markers were normal. We evaluated Collagen II markers because of the correlation

between DD and psoriasis[52] and the finding of abnormal Collagen II in psoriasis[53]. We

found no significant differences between groups in Collagen II synthesis marker PRO-C2_HP

and degradation marker C2M_HP or their ratios. The CALC2 synthesis marker was

undetectable in both cohorts, preventing the calculation of CALC2 synthesis-to-degradation

ratios.

Collagen III synthesis markers were normal, consistent with a prior report of normal circulating

Collagen III synthesis marker amino-terminal propeptide of type III procollagen (PIIINP) in DD

subjects[54]. Given the accumulation of Collagen III in DD, we expected to find elevated

Collagen III synthesis to degradation marker ratios. Instead, we found normal Collagen III

synthesis to degradation ratios. This finding may reflect that other proteases also degrade

Collagen III, and we did not measure all potential collagen degradation fragments. To

investigate further, we assayed a different Col III degradation marker, CTX-III_HP, on the

remaining aliquots. Only 17 (10 DD, 7 Controls) of the initial 27 (13 DD, 14 Controls) leftover

aliquots had sufficient residual volumes for this assay. Although this small subgroup's CTX-III

violin plot appearance suggested reduced degradation in DD compared to controls

(Supplementary Fig. S5), this difference was not statistically significant.

We found significantly lower Collagen V synthesis marker Pro-C5 levels in the DD cohort than in

controls (Figure 1B). This finding is unexplained. An increased percentage of Collagen V is

reported in DD tissues[33], suggesting a high Collagen V synthesis to degradation ratio.

However, we could not calculate this ratio because a Collagen V degradation marker was

unavailable.

The DD cohort also had significantly lower Collagen VII synthesis marker Pro-C7 than controls

(Figure 1C). We could not calculate Collagen VII synthesis to degradation ratios because

Collagen VII degradation markers were undetectable in both cohorts. Although there are no

reports of Collagen VII in DD, its role in matrix mechanobiology and elevated tissue levels in

systemic sclerosis[55] make it a potential marker for collagen metabolism abnormalities in DD.

We found no significant differences in the measurements of the remaining collagen markers. We

measured Collagen IV markers because Collagen IV is increased in DD tissues[16]. We

assayed Collagen VI markers and ratios because prior pathway analysis suggests a role for

Collagen VI in DD and other fibrotic diseases[56]. We selected Collagen IX markers based on

unpublished data suggesting a role in DD. None of these markers or ratios showed significant

differences between disease and control cohorts.

Plasma proteins and interactions in DD

Separate from the Nordic collagen fragment analysis, we analyzed whole proteins with the

SomaScan aptamer-based platform. The SomaScan panel measures 6995 protein binding sites

associated with 5444 proteins. Supplementary Table S6 lists genes and gene aliases of proteins

mentioned in this manuscript.

SomaScanc hypothesis-free analysis identified 54 distinct proteins with significantly different

plasma levels than controls. Twenty-four proteins had increased levels, and thirty had lower DD

levels than controls (Supplementary Table S1). Our findings are consistent with existing

literature.

The body of DD research publications mentions only seven of these 54 differentially expressed

proteins: Aggrecan (*ACAN*), Amine Oxidase Copper Containing 3 (*AOC3*), Caspase-3 (*CASP3*),

Periostin (*POSTN*), SMAD Family Member 1 (*SMAD1*), Spartin (*SPART*), and Ubiquitin

carboxyl-terminal hydrolase 8 (*USP8*). Our differential expression findings concur with reports of

ACAN[57], *AOC3*[13], and *POSTN*[58] overexpression in tissue and *SMAD1*[59], *SPART*[60],

and *USP8*[60] underexpression in cell culture. CASP3 levels were low in our findings, but are

upregulated in a report of transcriptomic profiling[57]. This finding is unexplained. *ACAN* is also

near the variant rs6496519 associated with DD[61]. Supplementary Table S1 summarizes the

relevance of these 54 genes to DD.

We constructed PPI networks and enrichment analysis of these 54 plasma proteins, revealing

significantly enriched protein-protein interactions in 32, with multiple enriched protein groups

and categories (Supplementary Fig. S2, Supplementary Table S2). These categories are

aligned with known Dupuytren cellular biology. Abnormalities in the KEGG category has04610

(complement and coagulation cascades) were consistent with DD pathway meta-analysis

findings[62]. Complement fragments induce adhesion molecules that act on endothelial cells,

consistent with DD-related perivascular clusters of immune cells and small vessel occlusion[13].

Complement fragments also activate mast cells that release TNF, a known DD pathway

component[63], and histamine, a driver of itching, which is common in early-stage DD[64]. Cell

membrane and extracellular matrix signaling receptor binding of growth factors, cytokines, and

other signaling factors play pivotal roles in DD pathobiology[14]. Protein phosphorylation

processes are consistent with increased phosphorylated STAT1, STAT3, SMAD2, SMAD3, and

ERK1/2 in DD tissues[59, 65, 66]. The enriched Reactome HSA-162582 (Signal Transduction)

pathway is noteworthy, as it includes signaling by TGF-Beta, WNT, and NOTCH, all prominent

in DD biology[67].

We based our hypothesis-free DPRS on high levels of Osteoclast-associated immunoglobulin-

like receptor (*OSCAR*) and low levels of Spartin (*SPART*) (Table 2). Our finding of elevated

OSCAR levels is interesting because Collagen I is an OSCAR ligand. OSCAR also stimulates T-

cells to release TNF[68], a component of DD biology[63]. Our finding of decreased *SPART*

levels is consistent with low *SPART* expression in DD fibroblasts[60]. SPART is also associated

with Epidermal Growth Factor Receptor (EGFR) degradation and transport to the cell

membrane[69], consistent with increased ratios of intracellular to cell membrane EGFR levels

found in DD[70].

Separate from the hypothesis-free analysis, we conducted a hypothesis-based analysis using a

subset of the SomaScan proteins selected from an independent literature review

(Supplementary Table S3). We found 23 proteins in this subset with significantly different

plasma levels than controls. Nine proteins had higher, and fourteen had lower levels in DD than

in controls. Nine proteins were present in both hypothesis-free and hypothesis-based groups.

Our findings are consistent with existing literature. Supplementary Table S4 summarizes the

potential relationships of these proteins and parent genes to DD.

With our hypothesis-free analysis, we performed a STRING PPI analysis of the hypothesis-

based findings to identify network relationships of the 23 proteins. We found significantly

enriched protein-protein interactions in 21 with multiple enriched categories (p-value = 1.35e-07,

Supplementary Table S5). This analysis implicated enrichment categories in addition to those

from the hypothesis-free analysis. Myofibroblast differentiation, proliferation, motility, WNT

signaling, and interleukin pathways are unsurprising, given their known roles in DD biology[10,

20]. Cancer pathway associations are also unsurprising, given the correlations of DC with

carcinomas and sarcomas[71, 72]. Infection-related pathways were unexpected and exclusively

involved viral and parasitic infection categories. This finding might be due to a hidden variable of

tissue repair biology, which overlaps interleukin-driven parasitic immune defense biology and

interleukin-related DD biology.

This hypothesis-based analysis provided data to generate a second Diagnostic Proteomic Risk

Score (DPRS) calculated on eleven proteins (Table 3). Figure 4A shows the AUC-ROC of this

panel (AUC = 0.9), and Figure 4B shows the DPRS based on this panel, distinguishing DD from

control cohorts.

We evaluated the potential of each DPRS for staging. Recurrence after a procedure is the

benchmark of Dupuytren's biological severity. Traditionally, surgeons use clinical factors such

as early age of onset, affected family members, and disease locations other than the palms to

predict post-procedure recurrence. However, the study's inclusion criteria included these

factors, so we could not use them to subset our Dupuytren cohort. Instead, we used age at the

time of the first DC procedure. Small studies suggest that recurrence is more common in

patients younger at their first DC procedure[73, 74]. We evaluated the ability of each DPRS to

distinguish Dupuytren subjects who had their first corrective procedure before age 50 versus

those having their first correction at age 50 or older. The hypothesis-based DPRS distinguished

DD subjects requiring their first DC procedure before age 50 from those having their first

procedure after age 50 (p=0.0018), suggesting potential use as a staging and predictive tool.

Although Mass Spectrometry did not reveal statistically significant differences in abundance

across cohorts, there was statistically significant concordance between Mass Spec and

SomaScan findings. The Mass Spec data contained 6 of the 54 proteins with statistically

significant differences between disease and control cohorts using the hypothesis-free

SomaScan data analysis: Amine Oxidase Copper Containing 3 *AOC3*, Periostin *POSTN*, Serpin

Family C Member 1 *SERPINC1*, Transferrin *TF*, Proprotein Convertase Subtilisin/Kexin Type 9

PCSK9, and Kininogen 1 *KNG1*. Table 4 compares the results of the SomaScan and Mass

Spectrometry data analysis. A pathway analysis of these six proteins identified two highly

significant Reactome categories (Supplementary Table S7). In addition to protein

phosphorylation noted above, Reactome HSA-381426 involving Insulin-like Growth Factor *IGF*

and Insulin-like Growth Factor Binding Proteins *IGFBP*s is relevant because of the role of IGF-II

and IGFBP-6 in regulating Dupuytren fibroblast proliferation and contractility[75].

This study has limitations. Statistical disadvantages were the small number of samples in all

assays and the large number of aptamers in the SomaScan assay. Although our findings align

with current research on Dupuytren proteomics, there are differences. Elevated serum

TIMP1[18] and TNF[20] have been reported in DD, but we found no statistically significant DE

between disease and control groups. This discrepancy may reflect that SomaScan assays

target specific protein conformations, whereas ELISA measures all conformations.

Conclusions

We used ELISA, aptamer, and mass spectrometry assays to identify differential plasma

proteomic signatures between DD and control cohorts. We found significant DD-related

proteomic abnormalities in both collagen metabolism peptide fragments and whole proteins in

circulation. This study is the first to have evidence suggesting that net collagen accumulation in

DD derives from normal collagen synthesis and impaired collagen degradation, challenging the

model of excess collagen synthesis and distinguishing DD from other organ fibroses. We

independently analyzed proteins from the same cohorts. Our hypothesis-free analysis of 5444

proteins revealed 54 with statistically significant differences between the DD and control

cohorts. Our independent hypothesis-based analysis of 328 proteins selected from the literature

review identified 23 with statistically significant differences. We developed two Dupuytren

Diagnostic Proteomic Risk Scores separately based on hypothesis-free and hypothesis-driven

analyses. We used pathway analyses to identify DD protein-protein network enrichments,

confirming known DD interactions and revealing previously unreported DD pathway categories.

Finally, we showed our hypothesis-based DPRS correlated with the subject's age at the time of

the first DD procedure, an index of biological severity. These findings suggest that a validated

hypothesis-based DPRS could guide management and develop DD therapeutics.

We recommend a larger study to validate our Collagen I, III, V, and VII metabolite findings. We

also recommend a validating study of 68 unique proteins identified in the protein aptamer

analysis (54 in the hypothesis-free analysis and 23 in the hypothesis-based analysis, with 9 in

both groups). Testing only the subset of the 68 most pertinent proteins would increase the

statistical power. Individual ELISA assays would be cost-prohibitive. An alternative approach

would be to develop a custom multiplex panel for the 13 proteins in our two DPRS[14] to

validate our DPRS and the correlation of our hypothesis-based DPRS on age at the first DD

procedure. Our findings strongly support the need to fund and conduct more extensive

validating studies of circulating collagen metabolites and proteins in DD. This biomarker

research is fundamental to developing DD diagnostic tests to diagnose, stage, and inform

research to develop therapeutic drugs for DD.

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Author Contributions

CE provided the study design and wrote the main manuscript text. CE and BH wrote the results

section, and BS and PS wrote the methods section. All authors performed data analysis.

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Supporting information

Data availability

All the data and analytic scripts that support the findings of this study are openly available in the

GitHub repository Dup_Blood_Proteomic_Data at

https://github.com/Dupuytren/Dup_Blood_Proteomic_Data.

Competing interest statement

868   These authors declare the following competing interests: Morten Karsdal, Cecilie Bager, and

869   Signe Holm Nielsen are all currently employed by Nordic Bioscience; Morten Karsdal, Cecilie

870   Bager, and Signe Holm Nielsen hold stocks in Nordic Bioscience. The rest of the authors have

871   no conflicts of interest to declare.

Figure sources

Figure 1. Each panel image was generated in R https://www.r-project.org. The final image was

assembled and annotated with Moho 13.5.5 https://moho.lostmarble.com.

Figure 2. The image was generated in R https://www.r-project.org.

Figure 3. Panels a and b were generated in R https://www.r-project.org. Panel C was

downloaded from STRING https://string-db.org. The final image was assembled and annotated

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Figure 4. Panels a and b were generated in R https://www.r-project.org. Panel C was

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with Moho 13.5.5 https://moho.lostmarble.com.

Supplementary Fig. S1 was created in PowerPoint.

Supplementary Fig. S2 was created in Excel.

Supplementary Fig. S3 was created in Excel.

Supplementary Fig. S4 was generated in R https://www.r-project.org.

Supplementary Fig. S5 was generated in R https://www.r-project.org.

Figure legends

Figure 1. Differential production of Dupuytren vs. control collagen metabolism markers. A)

Collagen I degradation marker C1M. B) Collagen I synthesis to degradation marker ratio Pro-

C1/C1M. C) Collagen V synthesis marker Pro-C5. D) Collagen VII synthesis marker Pro-C7. * p-

value <0.05, **p-value <0.005.

Figure 2. Heatmap of log-transformed standardized values of the 54 significant proteins in the

hypothesis-free analysis.

Figure 3. Hypothesis-Free DPRS and Protein-Protein Interactions. A) The Hypothesis-free

Diagnostic Proteomic Risk Score has an AUC-ROC value of 0.8. B) Hypothesis-free Diagnostic

Proteomic Risk Scores (HF-DPRS) of the control vs Dupuytren cohorts. p-value = 5.28E-06. C)

Network expression of 54 differentially expressed proteins. Of the 54 proteins, 30 had strong

pathway connections in the network: ACAN, AKR1A1, C5, C6, CALCB, CASP3, CGA, CGB3,

CPB1, CRKL, DSG2, EIF4H, FKBP5, G3BP2, GP1BB, GUSB, HGS, HSP90AA1, KNG1,

MAP3K11, MBNL1, MBNL2, NRG4, OSCAR, PCSK9, POSTN, PRSS1, SERPINC1, SOCS3,

SYK, TF, USP8, and YWHAB.

Figure 4. Hypothesis-based DPRS and Protein-protein interactions. A) The Hypothesis-based

Diagnostic Proteomic Risk Score has an AUC-ROC value of 0.9. AUC-ROC (Area Under the

Curve - Receiver Operating Characteristics) is an index of predictive accuracy. B) Hypothesis-

free Diagnostic Proteomic Risk Scores (HF-DPRS) of the control vs Dupuytren cohorts. (p-value

= 7.79E-09). C) Network expression of 23 differentially expressed proteins. Of the 23 proteins,

20 had strong pathway connections: DAB2, SMAD1, USP8, PLAT, SFRP4, ACAN, DDR2, HPX,

KNG1, LCN2, SERPINH1, TF, MAP2K1, PRKCA, STAT1, YWHAZ, POSTN, COL1A1, CASP3,

and STAT3.

Figure 1.

Figure 2.

Figure 3.

Figure 4.

C

Table 1. Collagen epitope markers and findings

Assay	D/C	Description	Represents	p-value	SE	FC (D/C)
PRO_C1_HP	ns	Internal epitope in the N-terminal propeptide of type I collagen	Col I Synthesis	0.9437	C 10.99 D 11.89	0.93
C1M_HP	Down**	MMP-generated fragment of the a1 chain of type I collagen	Col I Degradation	0.0036	C 5.25 D 4.30	0.73
PRO-C2_HP	ns	N-terminal fragment of type IIB procollagen	Col II Synthesis	0.7808	C 1.45 D 3.11	1.01
CALC2	ND	C-terminal propeptide of type II collagen	Col II Synthesis	NA	NA	NA
C2M_HP	ns	MMP-generated fragment of type II collagen	Col II Degradation	0.1607	C 1.52 D 1.16	1.29
PRO-C3_roHP	ns	N-terminal propeptide of type III collagen	Col III Synthesis	0.9951	C 1.04 D 1.24	0.93
C3M_HP	ns	MMP-9 generated fragments of type III collagen	Col III Degradation	0.1615	C 16.96 D 13.61	0.86
CTX-III_HP	ns	MMP-generated fragment of crosslinked type III collagen	Col III Degradation	0.4887	C 40.88 D 53.26	0.91
PRO-C4_HP	ns	Internal epitope in the 7S domain of type IV collagen	Col IV Synthesis	0.0739	C 195.98 D 183.12	0.84
C4M_HP	ns	MMP-generated fragment of the a1 chain of type IV collagen	Col IV Degradation	0.1014	C 8.89 D 7.42	0.84
PRO-C5_HP	Down*	C-terminal propeptide of type V collagen	Col V Synthesis	0.0458	C 33.17 D 35.88	0.82
PRO-C6_roHP	ns	C-terminal propeptide of type VI collagen alpha-3 chain	Col VI Synthesis	0.7222	C 0.27 D 0.27	0.92
C6M_HP	ns	MMP2-generated fragment of type VI collagen	Col VI Degradation	0.1097	C 22.31 D 23.41	0.83
PRO_C7 HP	Down*	BMP-I mediated neo-epitope in the NC-2 propeptide of type VII collagen	Col VII Synthesis	0.0480	C 2.49 D 1.08	0.72
C7M	ND	Neo-epitope of MMP13 degradation of type VII collagen	Col VII Degradation	NA	NA	NA
PRO-C9	ns	C-terminus of NC-I domain of the a1 chain of type IX collagen	Col IX Synthesis	0.9658	C 0.80 D 0.65	0.92

Synthesis and degradation markers of Collagen I, II, III, IV, V, VII, and IX. Compared to controls, there were significant reductions in markers of Collagen I degradation, Collagen V synthesis, and Collagen VII synthesis. D/C: ratio of Dupuytren to control values. FC: fold change. **: <0.005, *: <0.05. ns: not significant. ND: undetectable in both disease and control cohorts.

Table 2. Hypothesis-free Diagnostic Proteomic Risk Score

Protein	Eff	SE	p-value	FDR
Spartin	-11.62125	2.97479	0.000094	0.18
Oscar	13.99037	3.596067	0.0001	0.18

Proteins of the Hypothesis-free DPRS. Eff = log (odds ratio comparing Dup to control) for a change of 1 in the log (protein data). SE: Standard error. FDR: False Discovery Rate (Adjusted p-value).

Table 3. Hypothesis-based Diagnostic Proteomic Risk Score

Protein	Eff	SE	p-value	FDR
SMAD1*	-6.3649	1.877804	0.0007	0.12
CSMD1	5.397535	1.907423	0.00466	0.18
SH3BP2	-4.59151	1.582341	0.00371	0.18
POSTN*	4.305245	1.502627	0.00417	0.18
ACAN*	5.589134	1.75746	0.00147	0.12
LCN2	-4.73895	1.734879	0.0063	0.19
AOC3*	7.050175	2.62144	0.00716	0.2
DDR2	1.63418	0.595212	0.00604	0.19
USP8*	-3.30095	1.041269	0.00152	0.12
CSNK1G2*	-10.3331	3.593633	0.00404	0.18
SERPINH1	-2.6813	0.799538	0.0008	0.12

Proteins of the Hypothesis-based DPRS. Eff = log (odds ratio comparing Dup to control) for a change of 1 in the log (protein data). SE: Standard error. FDR: False Discovery Rate (Adjusted p-value). * Five also had significant differential expression in the 54 proteins identified in the hypothesis-free analysis.

Table 4. Mass Spectrometry (MS) – SomaScan (SS) concordance

Protein	Est MS	p-val MS	FDR MS	Est SS	p-val SS	FDR SS
AOC3	0.214854	0.076413	0.232677	0.116624	0.000845	0.217972
POSTN	0.410212	0.098842	0.232677	0.182892	0.000312	0.211479
SERPINC1	0.071331	0.116338	0.232677	0.091474	0.001500	0.225251
TF	0.035200	0.567393	0.829196	0.076715	0.000025	0.172072
PCSK9	0.165811	0.755318	0.829196	0.185675	0.001371	0.222267
KNG1	0.010065	0.829196	0.829196	0.065561	0.000490	0.217972

Six proteins identified by MS were also in the 54 proteins found in the hypothesis-free SS analysis. AOC3, POSTN, SERPINC1, TF, PCSK9, and KNG1. Of these, AOC3, POSTN, and SERPINC1 had False Discovery Rate (FDR, Adjusted p-value) <0.25 in both Mass Spec and SomaScan analyses.