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2 **Reliability of protein abundance and synthesis measurements in**
3 **human skeletal muscle**

4 Kanchana Srisawat^{1#}, Katie Hesketh¹, Matt Cocks¹, Juliette Strauss¹, Ben J. Edwards¹, Paulo J. Lisboa²,
5 Sam Shepherd¹ and Jatin G. Burniston^{1†}

6

7 ¹Research Institute for Sport & Exercise Sciences, and ²Department of Applied Mathematics, Liverpool
8 John Moores University, Liverpool, L3 3AF, United Kingdom.

9 #Current address: Department of Diseases control, Ministry of Public Health, Nontaburi, 11000,
10 Thailand.'

11

12 †Address for Correspondence: Professor Jatin G. Burniston
13 Research Institute for Sport & Exercise Sciences
14 Liverpool John Moores University,
15 Tom Reilly Building, Byrom Street,
16 Liverpool, L3 3AF, UK
17 Tel: +44 (0) 151 904 6265
18 Email: j.burniston@ljmu.ac.uk

19 **Keywords:**

20 Deuterium oxide; heavy water; fractional synthesis rate; biosynthetic labelling; protein turnover;
21 proteome dynamics; repeatability; precision; coefficient of variation

22

23 **Abbreviations:**

24 ²H₂O, deuterium oxide; ABD, abundance; AU, arbitrary unit; CI, confidence interval; DPP, dynamic
25 proteome profiling; FSR, fractional synthesis rate; LFQ, label-free quantitation; RMA, reduced major
26 axis.

27

28 **Abstract**

29 We investigated the repeatability of dynamic proteome profiling (DPP), which is a novel technique for
30 measuring the relative abundance (ABD) and fractional synthesis rate (FSR) of proteins in humans. LC-
31 MS analysis was performed on muscle samples taken from male participants ($n = 4$) that consumed 4x
32 50 ml doses of deuterium oxide ($^2\text{H}_2\text{O}$) per day for 14 d. ABD was measured by label-free quantitation
33 and FSR was calculated from time-dependent changes in peptide mass isotopomer abundances. One-
34 hundred and one proteins had at least 1 unique peptide and were used in the assessment of protein
35 ABD. Fifty-four of these proteins met more stringent criteria and were used in the assessment of FSR
36 data. The median (M), lower- (Q_1) and upper-quartile (Q_3) values for protein FSR (%/d) were $M = 1.63$,
37 $Q_1 = 1.07$, $Q_3 = 3.24$. The technical CV of ABD data had a median value of 3.6 % (Q_1 1.7 % - Q_3 6.7 %),
38 whereas the median CV of FSR data was 10.1 % (Q_1 3.5 % - Q_3 16.5 %). These values compare favorably
39 against other assessments of technical repeatability of proteomics data, which often set a CV of 20 %
40 as the upper bound of acceptability.

41 Skeletal muscle is an accessible tissue in humans and offers a unique opportunity to study complex
42 human physiology, including ageing, polygenic disease and adaptations to exercise, which can be
43 challenging to reproduce in animal or cell models. Proteomic analysis of muscle is particularly relevant
44 because the proteome is the interface between gene-environment interactions that underpin the
45 current functional state of a tissue. Proteomic studies of human muscle have provided insight by
46 associating patterns of protein abundance or post-translational modification with different functional
47 states (reviewed in ^[1] amongst others). However, this static information does not capture dynamic
48 aspects of the proteome such as turnover or adaptation. Static information, even when collected in a
49 time-series, cannot give insight to protein turnover or the relative contributions that synthesis and
50 degradation make to changes in protein abundance. Proteins within human muscle exhibit a broad
51 range of different turnover rates and changes to both synthesis and degradation contribute to
52 adaptations in protein abundance ^[2]. We ^[3] recently developed the new technique of dynamic
53 proteome profiling (DPP) that can measure both the abundance and synthesis rate of individual
54 proteins in human muscle. DPP combines deuterium oxide (²H₂O)-labelling with peptide MS and offers
55 the first insight to dynamic aspects of the human proteome *in vivo*. To further establish DPP it is
56 important to investigate the reliability of the technique and estimate the sensitivity of DPP to detect
57 biologically meaningful changes in relative protein abundance (ABD) and fractional synthesis rate (FSR).

58 We report the repeatability of protein ABD and FSR data in replicate analysis of muscle samples from 4
59 sedentary men (age = 38 ± 7 y; body mass = 76 ± 4 Kg). Each volunteer gave their informed consent to
60 the experimental procedures, which were approved (16/WM/0296) by the Black Country NHS Research
61 Ethics Committee (West Midlands, UK) and conformed with the Declaration of Helsinki. Stable isotope
62 labelling of newly synthesised proteins *in vivo* was achieved by oral consumption of ²H₂O over a 14-day
63 period. Consistent with our previous work ^[3], participants consumed 50 ml of 99.8 atom % of ²H₂O four
64 times per day. Venous blood was collected bi-daily, and muscle was collected at baseline (day 0), and
65 after 4, 9, and 14 days of ²H₂O consumption. Samples (~100 mg) of vastus lateralis were taken using
66 the conchotome technique after administration of local anaesthetic (0.5 % Marcaine). Two biopsies
67 were taken from each leg in alternate order and all samples were obtained after an overnight fast.

68 Body water enrichment of ²H was measured in plasma samples against external standards by gas
69 chromatography-mass spectrometry ^[4]. Soluble proteins were extracted from muscle samples as
70 previously described ^[3]. Tryptic digestion was performed using filter-aided sample preparation ^[5].
71 Digests containing 4 µg of peptides were de-salted using C₁₈ Zip-tips (Millipore) and analysed by LC-MS
72 consisting of nanoscale reverse-phase ultra-performance LC (NanoAcquity; Waters Corp., Milford, MA)
73 and online ESI QTOF MS/MS (Q-TOF Premier; Waters Corp.). Samples (5 µl corresponding to 1 µg tryptic
74 peptides) were loaded by partial-loop injection on to a 180 µm ID x 20 mm long 100 Å, 5 µm BEH C₁₈

75 Symmetry trap column (Waters Corp.) at flow rate of 5 $\mu\text{l}/\text{min}$ for 3 min in 2.5 % (v/v) ACN, 0.1% (v/v)
76 FA. Separation was conducted at 35 $^{\circ}\text{C}$ via a 75 μm ID x 250 mm long 130 \AA , 1.7 μm BEH C_{18} analytical
77 reverse-phase column (Waters Corp.). Peptides were eluted using a linear gradient that rose to 37.5 %
78 ACN 0.1% (v/v) FA over 75 min at a flow rate of 300 nl/min . Eluted peptides were sprayed directly in
79 to the MS via a NanoLock Spray source and Picotip emitter (New Objective, Woburn, MA). Additionally,
80 a LockMass reference (100 $\text{fmol}/\mu\text{l}$ Glu-1-fibrinopeptide B) was delivered to the NanoLock Spray source
81 of the MS and was sampled at 240 s intervals. For all measurements, the MS was operated in positive
82 ESI mode at a resolution of 10,000 FWHM. Before analysis, the TOF analyser was calibrated using
83 fragment ions of [Glu-1]-fibrinopeptide B from m/z 50 to 1990. Peptide MS were recorded between
84 350 and 1600 m/z and muscle samples were analysed in duplicate, in a randomized order interspersed
85 by inter-sample blanks (5 μl 0.1 % FA separated over a 15 min linear gradient). Data-dependent MS/MS
86 spectra were collected from baseline (day 0) samples over the range 50–2000 m/z . The 5 most
87 abundant precursor ions of charge 2+ 3+ or 4+ were selected for fragmentation using an elevated (20–
88 40 eV) collision energy. A 30-s dynamic exclusion window was used to avoid repeated selection of
89 peptides for MS/MS.

90 Label-free quantitation (LFQ) was performed using Progenesis Quantitative Informatics for Proteomics
91 (Waters Corp.) consistent with our previous work (e.g. ^[3, 6, 7]). Analytical data were LockMass corrected
92 using the doubly-charged monoisotopic ion of the Glu-1- fibrinopeptide B and prominent ion features
93 were used as vectors to warp each data set to a common reference chromatogram. MS/MS spectra
94 were searched against the Swiss-Prot database (2018.7) restricted to Homo-sapiens (20,272
95 sequences) using a locally implemented Mascot server (v.2.2.03; www.matrixscience.com). Enzyme
96 specificity was trypsin (allowing 1 missed cleavage), carbamidomethyl modification of cysteine (fixed),
97 and m/z errors of 0.3 Da.

98 Log-transformed MS data were normalized by inter-sample abundance ratio, and protein relative
99 abundance (ABD) was calculated using nonconflicting peptides. Mass isotopomer abundance data were
100 extracted from MS spectra using Progenesis Quantitative Informatics (Waters Corp.). The abundance
101 of m_0 – m_4 mass isotopomers was collected over the entire chromatographic peak for nonconflicting
102 peptides used in LFQ. Incorporation of ^2H into protein is associated with a decrease in the molar fraction
103 of the monoisotopic (m_0) peak. Changes in mass isotopomer distribution follow a nonlinear bi-
104 exponential pattern as a result of the rise-to-plateau kinetics in ^2H enrichment of the body water
105 compartment, and the rise-to-plateau kinetics of ^2H -labelled amino acids in to newly synthesised
106 protein. Data were fitted using the Nelder-Mead simplex method to optimise for the rate of change in
107 the molar fraction of the m_0 peak. The rate constant (k) of change in mass isotopomer distribution is a
108 function of the number of exchangeable H sites and was accounted for by referencing peptide

109 sequences against the ^2H enrichment of amino acids in humans^[8]. Peptides were selected for statistical
110 analysis if they were (i) unique to a protein, (ii) fitted well ($R^2 > 0.85$) to the biexponential model, and
111 (iii) were detected in each technical replicate across the entire time series (0, 4, 9 and 14 d) of samples
112 from all 4 participants. Protein FSR was derived from the median k of peptides assigned to the protein.
113 All data processing and statistical analyses were performed in *R* version 3.5.2.

114 The enrichment of ^2H in body water increased at a rate of 0.135 ± 0.005 %/d and reached a peak of
115 2.14 ± 0.08 % on day 14. LFQ encompassed 101 proteins that had at least 1 unique peptide and ABD
116 ranged from 7.75 e^{+01} AU (CISY: citrate synthase) to 2.86 e^{+05} AU (HBA: hemoglobin subunit alpha). Fifty-
117 four proteins passed the more stringent filtering necessary for synthesis calculations and FSR ranged
118 from 0.37 %/d (CASQ1: calsequestrin-1) to 12.90 %/d (APOA1: apolipoprotein A-I). The median (M),
119 lower quartile (Q_1) and upper quartile (Q_3) values for protein FSR (%/d) were $M = 1.63$, $Q_1 = 1.07$, $Q_3 =$
120 3.24 .

121 The overall repeatability of proteome ABD and FSR data was assessed using reduced major axis (RMA)
122 regression that is appropriate when random error is expected in both x and y variables. RMA is
123 equivalent to ordinary least products regression^[9] and can distinguish between fixed- and proportional-
124 bias. Strong linear relationships (Figure 1 A and D) existed between replicate analyses of ABD
125 ($R^2=0.9989$; $p = 0.0104$) and FSR ($R^2=0.9535$; $p = 0.0104$). The 95 % confidence interval (CI) of the
126 intercept and slope were used to assess fixed- and proportional-bias, respectively. ABD data did not
127 exhibit fixed bias (95 % CI of intercept = $-68.2 - 38.7$) but there was evidence of proportional bias
128 because the 95 % CI for the slope ($0.951 - 0.957$) did not include 1. The 95 % CI for the slope of FSR
129 data ($0.9285-1.001$) did span 1 but the 95 % CI for the intercept of FSR data ($0.100 - 0.308$) did not
130 span zero, which suggests fixed bias exists between replicate analyses. RMA analysis summarises the
131 overall reliability of the proteomic data but each protein exhibits unique technical repeatability. CV is
132 representative of 68 % (1SD) of the likely variation in data and has been a commonly used (e.g. ^[6, 10, 11])
133 index in reliability studies. CV was used to illustrate relative differences in the technical repeatability of
134 ABD and FSR data on a protein-by-protein basis (Figure 1B and 1E, respectively). The mean CV of ABD
135 data was 5.5 % (SD = 6.6 %); $M = 3.6$ % (Q_1 1.7 % - Q_3 6.7 %), the mean CV of FSR data was 14.1 % (SD
136 = 13.6 %); $M = 10.1$ % (Q_1 3.5 % - Q_3 16.5 %). To give insight to biological variability amongst participants,
137 FSR and ABD data were filtered to exclude proteins with a technical repeatability of >5 %CV. The
138 biological variability of FSR (Figure 1F) was $M = 30.4$ % (Q_1 17.9 % - Q_3 42.0 %) and was approximately
139 double that of ABD data (Figure 1C; $M = 14.6$ %; Q_1 7.7 % - Q_3 25.1 %). Ordinary least squares regression
140 found that neither protein ABD, FSR or the number of peptides per protein predicted the level of
141 technical repeatability in FSR (Figure 1 G, H and I). Accordingly, there was no difference ($p=0.7511$)
142 between the CV of FSR calculated from single peptides versus proteins that had 2 or more peptides.

143 Lastly, the ability of DPP to replicate a given result (i.e. precision) was defined according to ISO 5725 in
144 which the precision of a method is denoted by its repeatability (r)

$$145 \quad r = 1.96 \sqrt{2Sw^2}$$

146 where Sw is the within-subject standard deviation. The practical interpretation of r is “the value below
147 which the difference between two measurements would lie with a probability of 0.95” [12]. Glycolysis
148 was the most dominant functional group amongst the proteins surveyed and the precision of ABD and
149 FSR measurements for enzymes of the glycolytic pathway in human muscle is summarised in Figure 2.
150 Hexokinase was the only enzyme of the glycolytic pathway that was not detected in the current analysis,
151 or in our previous work [6]. This may be an artifact of our standard fractionation procedure or it may
152 relate to the reported differences in subcellular location of hexokinase [13] compared to other glycolytic
153 enzymes [14]. Consistent with [15, 16] enzymes of the second phase of glycolysis were of greater ABD than
154 those belonging to stage 1. Interestingly, the opposite pattern emerged in FSR data and generally the
155 FSR of stage 2 glycolytic enzymes was less than those involved in stage 1.

156 An understanding of measurement precision is an important aspect of scientific investigation and is
157 prerequisite to proper experimental design. The average CV in ABD data presented here (6 %) compares
158 favorably with 12 % for LFQ in yeast [10], 6 % for SWATH-MS analysis in HEK293 cells [11] and 7 % in our
159 [6] previous assessment of in rat skeletal muscle. Methods for studying the dynamic aspects of the
160 proteome, particularly in humans *in vivo*, are less widely reported than static abundance data. The
161 average technical repeatability (14 %) of protein FSR was remarkably similar to the performance of the
162 more established techniques for measuring relative protein abundance. Especially so, because FSR
163 calculations require time-series analysis encompassing measurements of both precursor enrichment
164 and incorporation of label in to newly synthesised protein. Biological variability in protein ABD and FSR
165 was estimated using proteins with the highest levels of technical repeatability (CV <5 %). The biological
166 variability in FSR was approximately double that of protein ABD. Sample size calculations [17], based on
167 Q_3 biological variation, estimate a required n of 6 (ABD) or 15 (FSR) to detect a within-subject change
168 of 50 % ($\alpha = 0.05$, 80 % power). DPP of human muscle responses to resistance exercise [3] reported
169 changes in FSR that, generally, were of twice the magnitude of changes in ABD. The above observations
170 suggest DPP has an equal ability to detect changes in ABD and FSR in the setting of human exercise
171 physiology.

172 The current assessment of DPP was limited to a subset of the most abundant muscle cytosolic
173 proteins [16]. While repeatability of protein ABD was similar to more extensive profiling of rat muscle

174 ^[6], it remains to be shown whether yet deeper analysis of the muscle proteome would achieve
175 similar levels of technical repeatability in FSR data. Herein, we report high levels of repeatability for
176 the measurement of both ABD and FSR using peptides that were consistently resolved and detected
177 in all samples. However, at a confidence level of 95 %, fixed bias was detected between duplicate
178 analyses of FSR data and proportional bias was detected in the measurement of ABD. In addition,
179 approximately 50 % of proteins were not consistently resolved in all samples and these data were
180 excluded prior to statistical analysis. Increasing the number of proteins eligible for statistical analysis
181 is key to the future development of the DPP technique and could be achieved through the use of
182 more modern instrumentation and orthogonal peptide separation techniques. Whereas, less
183 stringent handling of missing data, for example by inclusion of proteins detected in a subset of
184 participants or samples, should be avoided because this diminishes the technical repeatability of ABD
185 and FSR measurements.

186 In conclusion, DPP is a robust technique for the assessment of protein abundance and synthesis rates
187 in human skeletal muscle. Peptide MS analysis of ²H₂O-labelled samples is a burgeoning discipline that
188 has promise in bringing new insight to dynamic aspects of the proteome. The ability of DPP to report
189 abundance and synthesis data on a protein-by-protein basis in humans *in vivo* is unique. In the future
190 DPP of human muscle may help to uncover new information regarding the complex effects of exercise,
191 ageing or chronic diseases on the rate of turnover, as well as abundance, of muscle proteins.

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224 Figure Legends

225 **Figure 1 – Repeatability of protein abundance and synthesis measurements**

226 Duplicate analysis of (A) normalised protein abundance (ABD) measured by label-free quantitation and
227 (D) fractional synthesis rate (FSR) measured by deuterium oxide labelling *in vivo* and peptide mass
228 isotopomer analysis. Abundance data (AU) are reported for 101 proteins measured in each technical
229 replicate of $n = 4$ participants at experiment day 0. Synthesis data (%/d) are reported for 54 proteins
230 measured in technical replicates of $n = 4$ participants in time-series analysis of samples collected at 0,
231 4, 9 and 14 d of deuterium administration. The line of best fit was calculated by reduced major axis
232 regression and used to investigate fixed- or proportional-bias in ABD or FSR data.

233 Density plots (B and E) illustrate the distribution and median CV for technical replication of ABD and
234 FSR data. Biological variation (C and F) was illustrated after filtering protein lists to remove proteins
235 with a level of technical CV >5 %.

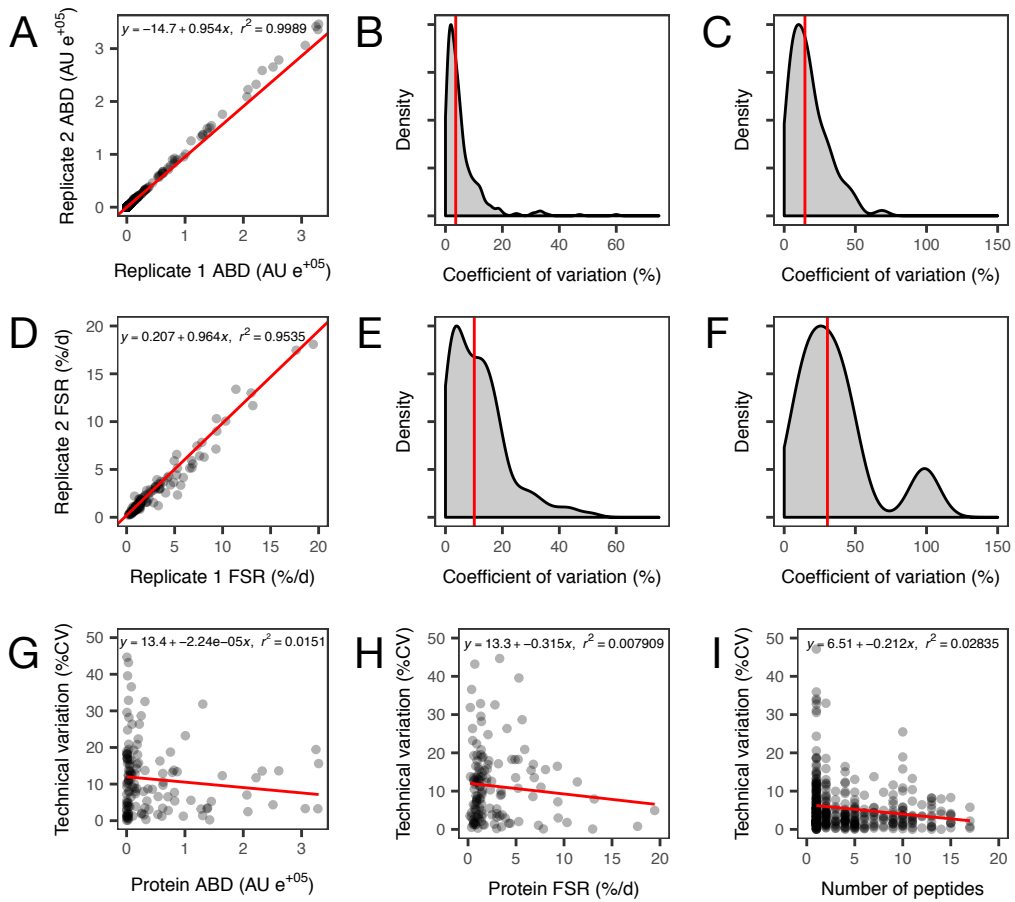
236 Scatter plots (G, H and I) illustrate the lack of significant relationship between technical variation in
237 protein FSR (%CV) and (G) protein ABD, (H) protein FSR, or (I) number of peptides per protein.

238 **Figure 2 – Dynamic proteome profiling of glycolytic enzymes in human skeletal muscle**

239 Gray boxes display the common name of each enzyme in the glycolytic pathway, redrawn from the
240 Kyoto Encyclopaedia of Genes and Genomes. Adjacent boxes detail the UniProt protein ID and number
241 of peptides. The median abundance (ABD) and fractional synthesis rate (FSR) of proteins is reported in
242 $n = 4$ participants. Data in parentheses represent the repeatability of ABD and FSR measurement
243 calculated according to ISO 5725 and defined as the maximum difference expected between two
244 measurements in 95 % of cases.

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