Mechanical loading induces distinct and shared responses in endothelial and muscle cells and reveals exercise-like molecular profiles

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Keywords

Acetate, Development, Metabolomics, PHGDH, Serine, Transcriptomics

Abstract

Skeletal muscles and blood vessels are continuously exposed to mechanical forces, particularly during exercise. We subjected human endothelial and skeletal muscle cells to cyclic mechanical stretch and investigated acute molecular responses. Mechanical loading elicited both shared and cell type-specific alterations in transcriptomic and metabolomic profiles, several of which mirror changes observed in vivo following exercise. Notably, both cell types released acetate in response to mechanical loading, however, many of the changes occurred in opposite directions. For example, genes associated with the electron transport chain were repressed in endothelial cells but upregulated in skeletal muscle cells. Furthermore, mechanical loading promoted a transcriptomic shift in endothelial cells indicative of maturation, and reduced their proliferation. Metabolic changes were more pronounced in endothelial cells, which exhibited increased serine biosynthesis from glucose, as demonstrated by ^{13}C-(U)-glucose tracing. Further experiments targeting phosphoglycerate dehydrogenase (PHGDH), a key enzyme in the serine synthesis pathway, underscored its role in endothelial cell anabolism. These findings suggest that mechanical loading alone can recapitulate several exercise-induced effects in endothelial and muscle cells, and highlight a potential link between mechanical stimuli, serine synthesis, and endothelial cell maturation.

Introduction

Skeletal muscle is a highly vascularized tissue and myofibers and endothelial cells together make up ~70% of the total nuclei count in skeletal muscle tissue (1, 2). Myofibers are contractile cells that both produce and sense mechanical forces. Endothelial cells lining the blood vessels sense mechanical forces produced by blood flow (shear stress) and by contracting and stretching muscle. In skeletal muscle, mechanical forces are produced and sensed especially during exercise, which can be mimicked *in vitro* using contraction and stretching models (3, 4). Previously, it has been demonstrated that cyclic mechanical loading (i.e. cyclic stretching) induces a wide variety of molecular and morphological responses in endothelial [reviewed in (5)] and muscle cells [reviewed in (6)]. For example, mechanical loading has been found to activate the mechanistic target of rapamycin complex 1 (mTORC1) (7) and to induce highly similar gene expression changes in murine myofiber-like multinucleated cells (myotubes) as resistance exercise *in vivo* in myofibers (8). Still, studies using human muscle cells are needed (4). Endothelial cells, in turn, have been shown to migrate and change orientation in response to stretch, secrete angiocrines, and activate angiogenesis-related pathways (5, 9, 10), which also occur in response to exercise and increased blood flow *in vivo* (11, 12). However, there are only a few studies (9, 13, 14) investigating the remodeling of the global molecular landscape in response to mechanical loading, and none in comparing the responses in different cell types. It is recently acknowledged that endothelial and muscle cell cross-talk is essential in regulating the development, maintenance, and repair of muscle tissue [reviewed in (2)].

Skeletal muscle responses to mechanical loading *in vivo* have mostly been examined using mixed muscle samples, which lack cell-specificity as skeletal muscle contains over 10 different cell types (1). *In vitro* models allow us to study metabolic responses in a cell type-specific manner. However, according to our knowledge, there are no studies in which the mechanical loading-induced responses of muscle and endothelial cells have been compared. Exercise induces many metabolic changes, such as increased glycolysis and oxidative phosphorylation at the whole-muscle level (15). However, little is known about how exercise or mechanical loading *per se* affects other metabolic pathways, such as those branching from glycolysis (e.g., the pentose phosphate and serine synthesis pathways), and their role in exercise responses such as angiogenesis. Previously, mechanical stretching of liver endothelial cells during vasodilation was demonstrated to induce angiocrine signals that support liver growth and survival, and potentially contribute to liver regeneration (10). Blood vessels in skeletal muscles experience immediate vasodilation—and thus endothelial cell stretching—when their physiological function is rapidly changed during exercise.

Here we explored how mechanical loading affects the transcriptional landscape and metabolomics of two different human endothelial cell lines, and primary human skeletal muscle cells, to mimic exercise on-a-dish and to study cell type-specific responses. Molecular responses resembled changes typically observed after exercise *in vivo*, however, many pathways were changed in opposite directions in endothelial cells and

87 muscle cells, demonstrating highly cell type-specific responses. The current exercise-
88 on-a-dish model induced acetate secretion from both cell types, suggesting its potential
89 role as an exerkine.

Results

Mechanical loading induced cell type-specific transcriptomic changes mirroring exercise-induced responses

We first investigated the transcriptomic changes in two human endothelial cell lines
(HUVEC & HCMEC) as well as human skeletal muscle cells (HSkM) (Fig. 1A). The used
mechanical loading scheme did not induce detachment of the cells, significant
membrane damage, or cell death in either muscle or endothelial cells as determined by
total RNA yield and the extracellular lactate dehydrogenase (LDH) activity assay
(Supplementary Fig. 1A-B). Differential gene expression analysis demonstrated that all
three cell types presented significant changes in their transcriptomes in response to the
mechanical loading, as demonstrated by the principal component analysis (PCA) (Fig.
1B). The transcriptomic response was the largest in HUVECs, with the total number of
differentially expressed genes being approximately 3-fold larger than in HCMECs or
HSkM cells (Fig. 1C). In HUVECs, 717 genes were significantly upregulated, and 779
genes significantly downregulated ($\log_2$FC ≥ 0.5 & adj. p ≤ 0.05), for HCMECs, 241 genes
were upregulated, and 261 genes were downregulated, and in HSkM cells 262 genes
were upregulated and 126 genes were downregulated. The ten most upregulated and
downregulated genes by indexing in order of magnitude and consistency of change
($\log_2$FC and adj. p) are annotated in Figure 1D for each cell type. For the complete lists
of differentially expressed genes and their full names, please refer to Supplementary
Table S1.

When comparing overlaps in both up- and downregulated genes in all three cell lines, we
noticed that the two endothelial cell lines clearly demonstrated similarities in their gene
expression, whereas the overlaps between endothelial cells and muscle cells were small
(Fig. 1E). Only one gene was upregulated (*CREB5*) and one gene was downregulated
(*FAM43A*) in all three cell lines in response to mechanical loading. For the complete lists
of overlapping differentially expressed genes, please refer to Supplementary Table S2.
These results demonstrated greater transcriptomic responses to mechanical loading in
endothelial cells than in muscle cells. Large overlaps in gene expression could be seen
between the endothelial cell lines, whereas endothelial cells and muscle cells
responded differently, highlighting the cell type-specificity.

Fig. 1 Changes in gene expression of human endothelial cells (HUVECs, HCMECs) and muscle cells (HSkM) in response to mechanical loading. **A)** Schematic representation of experimental

124 design. The experiment was repeated for each cell line separately. Cells from one stretching
125 chamber were used as one sample for RNA-Seq. **B)** PCA plots of samples in each cell line
126 respectively. **C)** The total number of detected genes and the number of genes that were either
127 up- or downregulated (log$_2$FC ≥ 0.5 and adj. p ≤ 0.05) for each cell line. **D)** Volcano plots
128 illustrating the amounts of differentially expressed and unchanged genes in all three cell lines.
129 Top 10 up- and downregulated genes by magnitude and consistency are annotated **E)** Venn
130 diagrams showing the number of shared up- and downregulated genes in all three cell lines

132 Gene Set Enrichment Analysis (GSEA) results revealed that the two endothelial cell lines
133 showed highly similar pathway enrichment after mechanical loading (Supplementary
134 Table S3). Twenty Gene Ontology: Biological Processes (GO:BP) and four Reactome
135 pathways were similarly upregulated in HUVEC and HCMEC, and 27 GO:BP, 83
136 Reactome and three KEGG pathways were similarly downregulated. The GSEA results of
137 HUVECs showed upregulation of cell maturation and EC junction-related pathways such
138 as *Endothelial cell development, Establishment of endothelial barrier, Cellular response*
139 *to nitrogen compound* and *Cell junction organization* (Fig. 2A). Furthermore, out of the
140 GO:BP *Endothelial cell development* -pathway, 18/72 genes were significantly
141 upregulated in HUVECs in response to mechanical loading (Fig. 2B). For muscle cells,
142 upregulated pathways were mostly related to oxidative phosphorylation (OXPHOS) and
143 ribosome biogenesis (Fig. 2D). Many of the GSEA findings were also reproduced by the
144 gene set overrepresentation analysis (ORA) (Supplementary Fig. 2).

145 To evaluate the external validity of our exercise-on-a-dish model, we analyzed the
146 expression of common exercise-induced and mechanosensitive genes. In endothelial
147 cells, we examined the expression of *KLF2/4, THDB, NRP2, YAP1, HIF1A*, and *NFE2L2*
148 and VEGF receptor 1 & 2 encoding genes *FLT1* and *KDR* (Fig. 2C), which have been shown
149 to be induced in ECs in response to flow-induced shear stress (16–18). Indeed, we found
150 that from these genes all but *HIF1A* were significantly upregulated in HUVECs in
151 response to the uniaxial mechanical loading. In HSkM cells, we specifically focused on
152 exercise-altered genes myostatin (*MSTN*), *NR4A3, VEGFA, MYC, ATF3,* and *TNFRSF12A*
153 (Fig. 2E), and observed similar changes in these genes after mechanical loading as have
154 been shown especially in response to resistance exercise in vivo (19–22). We
155 additionally looked at the expression of OXPHOS-related genes *PPARGC1A* and its
156 upstream regulator *PPARD* (23–25)*,* and found *PPARD* to be upregulated after
157 mechanical loading, while there was no change in *PPARGC1A* expression (Fig. 2E).
158 Together, these findings suggest that our *in vitro* mechanical loading model mimics
159 exercise and flow-induced mechanical stress in both endothelial and muscle cells.

160 When comparing gene set enrichment of the endothelial cell lines to muscle cells, we
161 found multiple pathways enriched in opposite directions in response to mechanical
162 loading (Supplementary Table S4). In HUVECs and HSkM cells 12 GO:BP, three
163 Reactome and two KEGG pathways were oppositely enriched and the same was true for
164 HCMECs and HSkM cells in 14 GO:BP, four Reactome and three KEGG pathways. Five of
165 these pathways were found to be oppositely regulated when comparing both endothelial
166 cell types to muscle cells. These pathways were related to cellular respiration,

metabolism, and chromatin remodeling (Fig. 2F). Most prominently, OXPHOS-related
pathways were downregulated in endothelial cell lines, while upregulated in muscle
cells. Closer inspection revealed that gene expression was specifically affected in the
same fashion throughout the electron transport chain (ETC)-complexes I-V (Fig. 2G).

Fig. 2 Transcriptomic changes of endothelial and muscle cells in response to mechanical
loading. **A)** Selected positively enriched gene sets in HUVECs from GO:BP, Reactome and KEGG
database terms. **B)** Upregulated individual genes in HUVECs in GO:BP *Endothelial cell*
development -pathway. **C)** Selected individual mechanosensitive endothelial genes in HUVECs.
D) Selected positively enriched gene sets in HSkM cells from GO:BP and Reactome database
terms. **E)** Selected individual exercise-altered genes in HSkM cells. **F)** Five gene sets from GO:BP,
Reactome and KEGG terms, that were significantly downregulated in both endothelial cell lines
and upregulated in muscle cells. **G)** Heatmap depicting expression of individual genes in all
electron transport chain complexes in response to mechanical loading. NES = Normalized
enrichment score. GO:BP = Gene ontology:Biological process. Rank = Rank of gene set by NES in
all statistically significant (adj. $p \leq 0.05$) up- or downregulated gene sets in their respective
database. Data are presented as mean ± SEM. *adj. $p < 0.05$, **adj. $p < 0.01$, ***adj. $p < 0.001$

Together, in endothelial cells mechanical loading induced positive enrichment of
pathways related to maturation and barrier establishment as well as upregulation of
known mechanosensitive genes. In muscle cells mechanical loading upregulated
OXPHOS and ribosome biogenesis pathways, as well as changes in typical resistance
exercise-regulated genes, suggesting that mechanical loading alone recapitulates
several common exercise effects seen *in vivo*. The transcriptomic results clearly
demonstrated cell type-specific responses induced by mechanical loading. This was
seen especially in the opposite expression of ETC-complex genes, which were
downregulated in endothelial cells and upregulated in muscle cells.

Mechanical loading resulted in greater metabolite changes in endothelial cells compared to muscle cells

Next, we compared metabolite changes in HUVECs and HSkM cells in response to
mechanical loading. We analyzed a targeted list of intracellular metabolites with UHPLC-
MS/MS, and culture media metabolites with non-targeted ¹H NMR spectroscopy (Fig. 3A,
Supplementary Table S5). In HSkM cells and HUVECs, 3/35 and 11/38 intracellular
metabolites were altered in response to mechanical loading, respectively (Fig. 3B). The
intracellular metabolite alterations were thus greater in HUVECs than in HSkM cells,
which is in line with the transcriptome results. From the extracellular metabolites
(conditioned media), 3/29 were altered in HSkM cells, and 2/34 in HUVECs (Fig. 3B).

By comparing conditioned media to the media incubated without cells (i.e., cell-free
media), we found that both HUVECs and HSkM cells released acetate into the media in
response to mechanical loading (Fig. 3C-D). We also confirmed the acetate release with
UHPLC-MS (Supplementary Fig. 3), which identified an even more pronounced acetate
release than ¹H NMR (Supplementary Fig. 3A-B). In addition, arginine and glutamine
levels were greater in the media of loaded HSkM cells and pyruvate levels in the media of
loaded HUVECs (Fig. 3C-D).

In muscle cells, intracellular levels of glucose 6-phosphate and phosphocreatine
decreased while alanine levels increased in response to mechanical loading (Fig. 3E). In

HUVECs, intracellular levels of phosphoenolpyruvate, fructose 6-phosphate, and α-
ketoglutarate decreased while phosphoserine, serine, asparagine, threonine, proline,
alanine, acetyl-CoA, and tryptophan levels increased (Fig. 3F). The two mostly increased
metabolites were phosphoserine and serine (Fig. 3F), whereas the serine synthesis
pathway gene expressions were unaffected (Supplementary Table S1). Proline and α-
ketoglutarate were the two most consistently (smallest *p*-values) changed metabolites.
The observed increase in proline and decrease in α-ketoglutarate may indicate that
proline was synthesized using α-ketoglutarate.

Next, we conducted a metabolite overrepresentation analysis using the significantly
altered intracellular metabolites. The overrepresentation analysis of HSkM cells did not
yield any significant results because of the low number of altered metabolites. In
HUVECs, the most enriched pathways were *Glycine and serine metabolism*, *Glutamate*
metabolism, and the *Warburg effect* (Fig. 3G).

Fig. 3 Intracellular and culture media metabolite changes of human skeletal muscle cells (HSkM) and HUVECs in response to mechanical loading. **A)** Culture media metabolites were analyzed with ^{1}H NMR spectroscopy, and intracellular metabolites were analyzed using the targeted

UHPLC-MS/MS method. Cells or media from three stretching chambers were pooled for one
sample. Also, media incubated without cells (i.e., Cell-free media) was included as an additional
reference. **B)** The number of increased, decreased, and unchanged intracellular and culture
media metabolites of HSkM cells and HUVECs. **C)** Significantly changed media metabolites of
HSkM cells. **D)** Significantly changed media metabolites of HUVECs. **E)** Significantly changed
intracellular metabolites in HSkM cells. **F)** Significantly changed intracellular metabolites in
HUVECs. **G)** Metabolite overrepresentation analysis using the significantly altered intracellular
metabolites of HUVECs. In panel G, the number of metabolites indicates how many significantly
altered metabolites from each pathway were detected. Individual values (n = 7-8 per group) are
plotted, and data are presented as mean ± SEM. The dashed lines represent the average
concentrations in cell-free media samples. P-creatine = Phosphocreatine, G6P = Glucose 6-
phosphate, PEP = Phosphoenolpyruvate, F6P = Fructose 6-phosphate
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Mechanical loading increased the synthesis of serine, alanine, and proline from glucose in endothelial cells

As serine metabolism was the most highlighted pathway in mechanically loaded
HUVECs, and its role in survival and growth has been demonstrated (26), we next
conducted dynamic labeling experiments with ^{13}C-(U)-glucose (Fig. 4A) to assess
whether the serine synthesis rate is affected by mechanical loading. We found that there
were significantly more total (unlabeled + labeled) and newly synthesized serine and
phosphoserine at multiple time points after mechanical loading (Fig. 4B-E). This
indicates that the serine synthesis was indeed increased in response to mechanical
loading in HUVECs. Total serine levels did not change in the media (Fig. 4F) whereas
mechanically loaded HUVECs released significantly more labeled serine into the culture
media (Fig. 4G), supporting the interpretation that serine synthesis was increased by
mechanical loading.

We also assessed the synthesis of alanine, proline, glycine, aspartate, asparagine,
glutamine, glutamate, and arginine. We observed more labeled and total alanine and
proline in loaded HUVECs and their media (Supplementary Fig. 4A-G), indicating that
alanine and proline synthesis were increased together with serine synthesis in response
to mechanical loading.

There was significantly less labeled glycine in loaded HUVECs at 3 h and 20 h time points
and total glycine at the 0 h time point (Supplementary Fig. 4H-I), but no differences in
labeled or total glycine release into media (Supplementary Fig. 4J-K) compared to non-
loaded controls. We did not find significant differences in labeled aspartate and
glutamate (Supplementary Fig. 5). In the case of asparagine, arginine, and glutamine,
^{13}C-labeled molecules were not detected after isotope natural abundance correction,
suggesting that HUVECs did not use glucose to produce these amino acids in these
nutritional conditions (Supplementary Table S6). Overall, the results suggest that
mechanical loading induces the synthesis of serine, alanine, and proline in endothelial
cells.

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Fig. 4 Mechanical loading increased glucose-derived carbon incorporation into serine and
phosphoserine in HUVECs. **A)** A schematic diagram of the time points of the ^{13}C-(U)-glucose
experiment. **B)** Total intracellular serine level (unlabeled + labeled). **C)** Newly synthesized (M+3)

serine level. **D)** Total intracellular phosphoserine (P-serine) level (unlabeled + labeled). **E)** Newly
synthesized (M+3) phosphoserine level. **F)** Total media (extracellular) serine level (unlabeled +
labeled). **G)** Newly synthesized (M+3) media serine level. Sample sizes were n = 3 per group per
time point in intracellular analyses and n = 5 in media analyses. Individual values are plotted, and
data are presented as mean ± SEM. *$p < 0.05$, **$p < 0.01$, ***$p < 0.001$.*
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Mechanical loading induces maturation of endothelial cells

Next, we tested whether mechanical loading alters the proliferation of endothelial cells.
We found that mechanical loading decreased proliferation of HUVECs, analyzed with
both flow cytometry- and microscopy-based assays (Fig. 5A-B), whereas overall protein
synthesis was not affected (Fig. 5C). These results, together with the transcriptomic
results for endothelial cell development (Fig. 2B), suggest that mechanical loading
induces maturation and junctional stability of endothelial cells and inhibits proliferation.
To translate our results to an *in vivo* situation, we applied the MetaMEx database (19) to
examine whether the *Endothelial Cell Development* -pathway is commonly induced in
response to acute aerobic exercise, and indeed we found this to be the case for several
of these genes (Supplementary Fig. 6).

Fig. 5 Effects of mechanical loading on endothelial cell proliferation and protein synthesis. A)
EdU-positive cells analyzed with flow cytometry (n = 3). B) EdU-positive cells in confluent and

sub-confluent cell densities were analyzed with confocal microscopy (n = 3). C) Protein synthesis
assessed by relative ^{14}C-phenylalanine incorporation (n = 6). **D**) Representative flow cytometry
histograms showing EdU negative and positive cell populations, and next to them are
representative confocal microscopy images showing all nuclei (dapi) and EdU positive nuclei.
Data are presented as mean ± SEM, and individual values are plotted when applicable. *$p < 0.05$.
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Serine synthesis pathway enzyme phosphoglycerate dehydrogenase (PHGDH) is critical for endothelial cell anabolism

Finally, to extend the previous findings on the role of the serine synthesis pathway in
endothelial cells (26), we used a lentiviral vector to overexpress PHGDH, the rate-
limiting enzyme of the serine synthesis pathway (Fig. 6A). Overexpression of PHGDH led
to a ~10% increase in protein synthesis ($p < 0.01$, Fig. 6A) and a trend for increased
proliferation ($p = 0.088$, Fig. 6C) in HUVECs. Next, we used chemical PHGDH inhibitors
and evaluated their effects on protein synthesis and proliferation. We discovered that
PHGDH inhibitors decreased the protein synthesis and proliferation of HUVECs (Fig. 6D-
E) without inducing cell death (Fig. 6F). We also found that the PHGDH inhibitors
decreased glucose-derived carbon incorporation into proteins, RNA, and lipids
(Supplementary Fig. 7). Overall, these data suggest that the serine synthesis pathway
enzyme PHGDH regulates endothelial cell growth, similarly to what we recently showed
in muscle cells (27, 28).

Figure 5. Effects of the serine synthesis pathway enzyme phosphoglycerate dehydrogenase
(PHGDH) manipulation on cell proliferation and protein synthesis. **A)** PHGDH level in response

to overexpression treatment (n = 16). **B)** Protein synthesis (^{14}C-phenylalanine incorporation into
proteins) in response to PHGDH overexpression treatment (n = 16, but in no treatment group n =
4). **C)** Cell proliferation in response to PHGDH overexpression treatment (n = 24, but in the no-
treatment group n = 6). **D)** protein synthesis (relative ^{14}C-valine incorporation into proteins) in
response to PHGDH inhibitors NCT-503 and CBR-5884 (n = 6). **E)** PHGDH inhibitors decreased
the proliferation of HUVECs, as assessed by EdU analysis. The groups were statistically
compared only to the DMSO vehicle control (n = 8 in the DMSO control group and n = 4 in the
inhibitor groups). **F)** PHGDH inhibitors did not induce cell death according to the lactate
dehydrogenase (LDH) activity measured from conditioned media (i.e., increased LDH activity
would indicate membrane damage and cell death) (n = 3). Individual values are plotted, and data
are presented as mean ± SEM. *$p < 0.05$, **$p < 0.01$, ***$p < 0.001$, **** $p < 0.0001$.*

Discussion

In this study, we unraveled the effects of mechanical loading on skeletal muscle and
endothelial cells and identified both shared and cell type-specific responses. Notably,
both muscle and endothelial cells significantly increased acetate release into the media
and induced gene expression changes typically observed after exercise. However,
transcriptomic responses exhibited several contrasting patterns in endothelial and
muscle cells. Most strikingly, OXPHOS and electron transport chain (ETC) genes were
upregulated in muscle cells but downregulated in endothelial cells. Furthermore,
endothelial cells demonstrated more pronounced metabolic responses, particularly in
amino acid metabolism. In response to mechanical loading, endothelial cells decreased
proliferation and upregulated expression of genes related to maturation and junctional
stability. Furthermore, using a ^{13}C-(U)-glucose tracer, we demonstrated that mechanical
loading enhanced glucose-derived serine synthesis in endothelial cells. Overexpression
of the serine synthesis enzyme PHGDH promoted protein synthesis of endothelial cells,
whereas PHGDH inhibitors decreased proliferation and anabolic activity, highlighting the
role of de novo serine synthesis from glucose in endothelial cell growth.

Both muscle and endothelial cells released acetate into the media in response to
mechanical loading, which was observed with two different analysis methods.
Previously, we found that electrically evoked myotube contractions also induced acetate
release from murine myotubes (29). In humans, it has been found that acetate is
released into the bloodstream (30) and urine (31) after exercise. Moreover, serum
acetate concentrations have been found to be greater in physically active individuals
compared to inactive individuals (32). Therefore, acetate could act as a possible exerkine
(i.e. a signaling molecule released in response to exercise) released by not only muscle
cells but also endothelial cells. One of the most relevant functions of acetate is acting as
an immunomodulator and promoter of T-cell effector function and regulating
differentiation of CD4$^+$ T-cell helper cells (33–35). This may partly explain the effects of
exercise in enhancing immune system function and increasing CD4$^+$ T-helper cell count
in humans (36). This is particularly interesting, as factors secreted by CD4$^+$ T-cells seem
to be important in angiogenesis, myogenesis, and muscle regeneration (37, 38). In
addition, the potential source of acetate after mechanical loading is interesting, as

acetate can be produced from (A) acetyl-CoA by certain acyl-CoA thioesterases (39), (B)
from pyruvate by ketoacid dehydrogenases and reactive oxygen species (40), and (C) by
deacetylation of proteins (41). Of these, at least reactive oxygen species production is
induced in muscle by exercise (42). Thus, acetate's possible role as an exerkine and its
links to the immune system or other human physiological systems should be studied
further.

Endothelial cells and muscle cells also demonstrated different and even opposite
transcriptomic responses to mechanical loading, most prominently in OXPHOS-related
pathways, and specifically in ETC complex genes. In endothelial cells the expression of
these genes was downregulated, and it has previously been shown that they primarily
rely on glycolysis for ATP production despite having access to oxygen, and that vessel
sprouting *in vitro* is regulated by glycolysis (43). OXPHOS-related genes were on the
other hand upregulated in muscle cells, and skeletal muscle is known to upregulate
mitochondrial biogenesis and oxidative respiration in response to physiological stress,
such as exercise (15). One of the well-known mechanisms regulating these exercise
adaptations is the AMPK-PGC1α pathway (23). In our study, although PGC1α gene
expression was unchanged in muscle cells, we did observe an increase in *PPARD*
expression after mechanical loading. PPARD is upstream of PGC1α (24) and has been
linked to enhanced oxidative metabolism in skeletal muscle (25). Furthermore, muscle
cells also demonstrated an upregulation of the *NR4A3, ATF3, and TNFRSF12A*, which
have been shown to belong to the five most upregulated genes in response to resistance
exercise (19). For example, *NR4A3* is essential in glucose metabolism and anabolism,
such as ribosome biogenesis in primary muscle cells (20). In line with the upregulation
of *NR4A3*, we also observed increases in the *ribosome biogenesis* GO:BP pathway after
mechanical loading, which is a known response to resistance training (44). Moreover,
we saw a decrease in myostatin (*MSTN*) expression in muscle cells, which is a common
response after a single bout of resistance exercise (21, 22). These results demonstrate
that mechanical loading of human skeletal muscle cells recapitulates several of the
common effects of exercise seen *in vivo*.

In endothelial cells, transcriptomic results showed that mechanical loading induced
upregulation of genes related to endothelial cell development, migration, cell-cell
junctions, and endothelial barrier integrity. Many of the endothelial cell development
genes were also found to increase in response to acute aerobic exercise in human
skeletal muscle tissue, which is known to be highly vascularized (1, 2, 19). Furthermore,
our results showed upregulation of mechanosensitive genes commonly associated with
an atheroprotective endothelial phenotype, which are typically induced by flow-
mediated mechanical stress (16–18). Our findings point out that a similar effect was
seen after cyclic stretching, which is comparable to skeletal muscle movement. We also
found that genes related to VEGF response, including VEGF receptors 1 & 2 (*FLT1 & KDR*),
were upregulated in endothelial cells. Related to these, *VEGFA* expression was also
increased in mechanically loaded muscle cells, and it has been hypothesized that
skeletal muscle stem cells recruit endothelial cells via VEGF-A expression and this
interaction is important for muscle stem cell self-renewal (45). Recently, a study

investigating acute endothelial-specific transcriptional responses in mouse skeletal
muscle found that exercise induced activation of pathways related to endothelial cell
development, differentiation, and migration, as well as VEGF response (46), which align
well with our results and validates that mechanical loading of endothelial cells
recapitulates effects of exercise *in vivo*.

In endothelial cells, we observed upregulation of heme signaling- and endothelial cell
maturation-related genes in response to mechanical loading. Further, ^{13}C-glucose
tracing analysis revealed that the abundance and synthesis of serine increased after
mechanical loading. Moreover, the overexpression of the serine-synthesizing enzyme
PHGDH increased protein synthesis, while commonly used non-competitive PHGDH
inhibitors decreased proliferation, protein synthesis, and glucose utilization for
anabolism. This is similar to what we recently demonstrated in muscle cells (27, 28).
Loss of PHGDH has previously been shown to cause lethal defects in angiogenesis in
mice and knockdown of PHGDH impairs endothelial cell proliferation (26). Moreover,
PHGDH has been highlighted as essential for the regulation of heme production in
endothelial cells (26), and hemin (oxidized heme) supplementation promotes
differentiation of endothelial progenitor cells (47). These results suggest that
mechanical stimuli/exercise, the serine synthesis pathway, heme signaling, and
endothelial cell maturation may be interconnected.

Strengths and limitations

We used two different human endothelial cell types and primary human skeletal muscle
cells. The loading induced several common exercise-induced and mechanosensitive
genes, suggesting an external validity of the exercise-on-a-dish model. Moreover, several
of the transcriptomic responses to mechanical loading were similar between the two
endothelial cell types (HUVEC and HCMEC), suggesting that many of the changes can be
applied to endothelial cells in general. We also replicated the main metabolite and
proliferation results with different methods. For example, acetate release was observed
with both ^{1}H NMR spectroscopy and UHPLC-MS, and the decrease in proliferation was
determined with microscopy- and flow-cytometry-based analyses. In addition, the
increased serine, alanine, and proline content and synthesis were detected by targeted
metabolomics and isotopic tracing from multiple time points.

The study also has limitations. We conducted the experiments with monolayers of cells
(i.e., two-dimensionally), and therefore, myotubes were not aligned unidirectionally
while the mechanical loading was unidirectional. In contrast, smaller endothelial cells
could change their alignment against the direction of stretch. This difference in the ability
to change alignment may in part explain why we detected fewer molecular changes in
myotubes compared to endothelial cells, although myotubes also recapitulated several
exercise responses. Overall, the data suggest that the mechanical loading model for two-
dimensionally cultured cells is especially suitable for endothelial cells, which also grow
in monolayers *in vivo*, while myotubes may need to be loaded three-dimensionally to
sufficiently mimic exercise in future studies. The potential importance of the mechanical

loading and serine synthesis pathway for improved vascular function should be studied
in vivo in further studies.

Conclusion

Mechanical loading induces both common and highly cell type-specific transcriptomic
and metabolomic responses in endothelial and skeletal muscle cells. In general, several
of the effects recapitulated what is observed *in vivo* after exercise. Both endothelial and
muscle cells released acetate in response to mechanical loading, and therefore
acetate's potential role as an exerkine could be an interesting topic for future studies.
The responses to mechanical loading were larger in endothelial cells, revealing
increased serine synthesis and upregulation of endothelial cell development as one of
the main responses. These results highlight that mechanical stimulus, serine synthesis,
and endothelial cell maturation may be interconnected.

Materials and methods

Endothelial and muscle cells

We used human umbilical vein endothelial cells (HUVEC, Cat# C-12200, PromoCell),
human cardiac microvascular endothelial cells (HCMEC, Cat# C-12285, PromoCell),
and primary human skeletal muscle (HSkM) cells. The latter were obtained as previously
(27) from two healthy 25-34-year-old volunteers (1 male and 1 female). Vastus lateralis
muscle biopsies were obtained under local anesthesia with the Bergström biopsy
needle. The human study (48) was conducted in accordance with the Declaration of
Helsinki and was approved as part of a larger TraDeRe-study by the Ethics Committee of
the University of Jyväskylä (857/13.00.04.00/2021). All participants provided written
informed consent for inclusion prior to participating in the study.

Cell culture

The HUVEC and HCMEC cells were cultured in endothelial cell growth medium (EGM)
(Cat# C-22010, PromoCell) supplemented with EGM SupplementMix (Cat# C-39215,
PromoCell) and 1% (vol/vol) of 1000 U/ml penicillin-streptomycin (P/S, Cat# 11548876,
Gibco). HSkM cells were cultured in growth medium (GM) consisting of high glucose (4.5
g/l) Dulbecco's Modified Eagle Medium (DMEM, Cat# 10313021, Gibco) supplemented
with 20% (vol/vol) fetal bovine serum (FBS, Cat# 10270106, Gibco), 1% (vol/vol) of 1000
U/ml P/S and 1% (vol/vol) 200 mM L-glutamine (Cat# 11500626, Gibco).

After reaching over 80% confluency in culture flasks, the cells were seeded (~20,000
cells/cm^2) on polydimethylsiloxane (PDMS) chambers (Cat# STB-CH-04, STREX). PDMS
chambers were pre-coated with FBS for 24 h and with 0.2% gelatin for an additional 24 h.
The PDMS chambers were placed in 150 mm diameter Petri dishes, two chambers per
dish. Cells were grown on PDMS chambers until reaching over 80% confluency. After
that, HUVEC and HCMEC cells were mechanically loaded, as described in the next

section. In the case of HSkM cells, at 80% confluency the GM was replaced with
differentiation medium (DM) containing high glucose DMEM, 2% (vol/vol) horse serum
(HS, Cat# 26050088, Thermo Fisher Scientific), 1% P/S, 1% L-glutamine and 1% (vol/vol)
Insulin-Transferrin-Selenium-X supplement (ITS-X, Cat# 51500056, Gibco). To induce
myoblast fusion into myotubes, the cells were cultured in DM for 4-5 days and DM was
changed every 48 h. Cells were cultured at 37 ℃ in a 5% CO_2 atmosphere all the time.

Mechanical loading set-up

Mechanical loading of cultured cells was achieved with a commercial cell culture pacing
system (C-Pace EM 100, IonOptix) paired with a motor-powered mechanical stimulation
unit (C-Stretch 100, IonOptix). The C-Stretch mechanical stimulation unit consists of a
motor-driven tray that causes uniaxial elongation of the PDMS chambers inserted in the
device and exposes cells, that are attached to the bottom of the chamber, to mechanical
load.

As mentioned, cells were seeded onto 12 PDMS chambers for each stretch experiment,
with 6 chambers loaded into the C-Stretch unit in series, and 6 chambers left as static
controls. The stretch unit was thoroughly cleaned with 70% EtOH before use. Fresh
media was changed for all chambers prior to starting the experiment. The stretch unit
was then placed in an incubator with conditions of 37 ℃ and 5% CO_2 atmosphere along
with the 6 control chambers. A cyclic mechanical stretch with a frequency of 1 Hz and
12% elongation was applied for 5 h. For each one-second cycle, a trapezoidal waveform
of 0.4 s stretch, 0.1 s hold, 0.4 s return, and 0.1 s hold, was used to keep the velocity of
the deformation of the silicone chamber at a moderate speed.

Lactate dehydrogenase activity assay

510   Lactate dehydrogenase activity (LDH) was measured from conditioned media as an
511   indicator of possible membrane damage and cell death (49). The analysis was
512   performed as previously (27) using an LDH Activity Assay Kit (Cat# 981906, Thermo
513   Fisher Scientific).

RNA extraction and analyses

RNA isolation, library preparation and sequencing

RNA was extracted using a commercial column-based RNA isolation kit (NucleoSpin®
RNA, Cat# 740955.250, Macherey-Nagel) immediately after the 5-hour mechanical
loading. The PDMS chambers were washed with PBS, and cells were lysed by pipetting
back and forth vigorously with lysis solution (350 μl of guanidium thiocyanate lysis buffer
+ 3.5 μl of β-mercaptoethanol). Extraction was then continued according to the kit's
instructions. An additional centrifugation step of 30 s at 11 000 g was included at the end
to ensure complete removal of the wash buffer. Bound RNA was then eluted into a
volume of 60 μl of RNase-free water. The collected elution was then aliquoted and
samples were immediately transferred to -80 °C. RNA quality control, library preparation
and sample sequencing were performed at Novogene (Cambridge, United Kingdom),

utilizing an Illumina NovaSeq 6000 platform, producing 150bp paired-end reads. Filtering
of raw sequence data was also conducted by Novogene in the following way: (1)
removing reads containing adapters, (2) removing reads containing >10% of unsolvable
bases and (3) removing reads containing >50% of low quality bases (Qphred ≤ 5). Filtered
RNA sequencing data is available in the Gene Expression Omnibus (GEO) data repository
(accession number: GSE288064).

Read alignment, normalization and filtering

Reads were aligned to the human genome (build GRCh38.95) using the Bowtie2 software
(50), with alignment rates >80% for all the samples. The aligned and mapped reads were
later annotated and quantified using the Rsubread package in R/Bioconductor software
(51), with the human GRCh38.110 genome build as annotation reference. The obtained
gene counts were transformed to counts-per-million (CPM) using the edgeR package in
the R/Bioconductor software (52). Genes with a CPM < 1 in three samples or more in
each group (50%), were subsequently filtered out as described in (53). The resulting
expression data were finally normalized using the trimmed mean of M-values (TMM)
algorithm (54).

Differential gene expression and pathway analyses

Differential gene expression (DGE) analyses were implemented through the
R/Bioconductor limma-voom pipeline (55). The voom-function of the limma package
was initially executed to format the data appropriately (56), after which empirical
Bayesian statistics were computed to rank genes based on their evidence for differential
expression. To account for type I errors induced by multiple testing, the Benjamini-
Hochberg method was used to adjust the tests' p-values, retrieving an adjusted p-value
(adj. p). Genes with a $\log_2$FC ≥ 0.5 and adj. p ≤ 0.05 were regarded as differentially
expressed. For subsequent Gene Set Enrichment Analysis (GSEA) genes were ranked in
an unbiased way by multiplying the $\log_2$FC by the -log10 of the adj. p to consider both the
magnitude and consistency of the change in gene expression. GSEA was performed with
the clusterProfiler R-package (57) with Gene Ontology: Biological Process (GO:BP),
Reactome, and the Kyoto Encyclopedia of Genes and Genomes (KEGG) serving as
reference databases. Ontologies and pathways were considered significantly enriched if
their corresponding tests yielded an adj. p-value <0.05. In addition, gene set
overrepresentation analysis (ORA) of the significantly up- or downregulated genes
($\log_2$FC ≥ 0.5 & adj. p ≤ 0.05) was conducted using the ShinyGO web-application v0.81
(https://bioinformatics.sdstate.edu/go/) (58) from GO:BP pathways. Pathway minimum
size was set to 2 and maximum to 5000 and the FDR cutoff for significant pathway
enrichment was set to 0.05. All analyses implemented in R/Bioconductor included the
core software version 4.3.1 with the R-studio interface version 2024.04.2.

Meta-analysis of skeletal muscle response to exercise (MetaMEx)

564 We examined gene expression of the *Endothelial cell development* pathway
565 (GO:0001885, 72 genes, list checked on 10.06.2025) in response to acute aerobic
566 exercise   using   the   MetaMEx   browser-based   tool

567 (https://www.metamex.eu/app/metamex) (19). All the other settings were kept as a
568 default, except that only the healthy population was included.

Metabolite extraction and analyses

^{1}H NMR spectroscopy metabolomics

Metabolites from the media samples were extracted as previously (27, 29). Cell-free
medium samples were also included as a reference to assess the uptake and release of
metabolites. Dried samples were reconstituted with 200 µL of sodium phosphate buffer
(150 mM, pH = 7.4) in 99.8% deuterium oxide (Acros Organic; Thermo Fisher Scientific)
containing 0.5 mM 3-(trimethylsilyl)propanesulfonic-d$_6$ acid sodium salt (DSS-d$_6$; IS-2
Internal Standard, Chenomx Inc.) and transferred to round-bottom 3-mm OD NMR tubes
(Cat# 341-PP-7, Wilmad). Data collection was carried out at 25 °C using the Bruker
AVANCE III HD NMR spectrometer, operating at 800 MHz ^{1}H frequency equipped with a
cryogenically cooled ^{1}H, ^{13}C, ^{15}N triple-resonance probe head (Bruker Corporation). The
^{1}H nuclear magnetic resonance (^{1}H NMR) spectra were collected as previously described
(29). ^{1}H NMR spectra were processed, and metabolites were identified and quantified
with the Chenomx 10.0 software (Chenomx).

UHPLC-MS/MS targeted metabolomics

Intracellular metabolites were collected and extracted as previously (27) with minor
modifications. In brief, 200 µL of ice-cold 80% methanol was added to each stretching
chamber, and metabolites from three chambers were pooled for one sample. Dried
samples were reconstituted with 40 µl of 50-50 acetonitrile-water mixture (vol/vol)
before analysis. Metabolites were analyzed as previously (27). Briefly, metabolites were
separated with an Atlantis Premier BEH Z-HILIC UHPLC column (1.7 µm, 2.1 x 100 mm)
which was used with an Agilent 1290 Infinity ultra-high-performance liquid
chromatography (UHPLC) system. *Mobile phase A* was 15 mM ammonium bicarbonate
(pH = 9.0) in water, and *mobile phase B* was 15 mM ammonium bicarbonate (pH = 9.0) in
90% acetonitrile-10% water. The flow rate was 0.7 mL/min, the column oven
temperature was 40 °C, and the injection volume was 3 µl. The following gradient elution
was used: the initial concentration of 90% *B* was held for 1 min and then was linearly
decreased to 65% *B* over the next 5 min and was held for 1 min at 65% *B*. Then *B*% was
increased to 90% over the next 1 min and was held 2 min at 90% *B* before next injection.
The UHPLC system was coupled with an Agilent 6460 triple quadrupole mass
spectrometer operated with dynamic multiple reaction monitoring mode (dMRM) as
previously described (27). A mixture of target metabolites (concentrations between 10
and 50 µM) was injected eight times at the beginning of the analysis and after every 6 to
8 samples for quality control and metabolite identification purposes.

Acetate was analyzed from media samples using the above-described chromatographic
method and the UHPLC-MS instrument which was operated on negative polarization. For
acetate, selected ion monitoring [e.g., pseudo-MRM (59)] of *m/z* 59 was done. The
following parameters were used for the selected ion monitoring experiment: fragmentor:
80; collision energy: 1; cell accelerator voltage: 4. The other ion source parameters were

the same as previously reported (27). Retention time and linearity of the signal were
tested using 10, 20, 40, 80, and 160 μM acetate standards that were prepared using
sodium acetate (Cat# 106268, Merck). The acetate standard was also used for a spike-in
experiment.

Batch correction, normalization, and statistical analyses of metabolomics data

Targeted UHPLC-MS/MS data and ^{1}H NMR data were measured within two batches and
therefore, they were batch effect corrected using the EigenMS method on the
MetaboAnalyst 6.0 platform (https://www.metaboanalyst.ca/). In the case of UHPLC-
MS/MS, data collected with negative and positive modes were separately batch-
corrected before combining the data. Single missing values within different groups were
replaced with 1/5 of the minimum detected values. Metabolite
concentrations/abundances were $\log_2$ transformed for statistical testing with the
Student's t-test with significance set at $p < 0.05$ and fold change $> |1.1|$.

Metabolite overrepresentation analysis

623 MetaboAnalyst 6.0 platform was used for the metabolite overrepresentation analysis.
624 Names of the significantly altered metabolites were uploaded. Default settings were
625 used (SMPDB pathway and '*Only use metabolites sets containing at least 2 entries*'
626 option were used).

^{13}C-glucose labeling experiments and analyses

Glucose-free and phenol red-free endothelial cell medium (Cat# H1168GPF, Cell
Biologics) medium including the medium kit's supplements, was supplemented with 5.5
mM U-^{13}C$_6$-D-glucose (Cat# CLM-1396-1, Cambridge Isotope Laboratories, isotopic
purity 99%). This medium was changed to the cells just before the initiation of the
experiment. Intracellular metabolites were extracted with the same protocol as for
targeted UHPLC-MS/MS metabolomics, but this time two chambers were pooled instead
of three. Media metabolites were extracted with the same protocol as for ^{1}H NMR
metabolomics described above, except now 100 μl of media from each chamber was
mixed with 200 μl of methanol. Intracellular and medium metabolites were reconstituted
into 45 μl and 400 μl of 50-50 acetonitrile-water solution, respectively. Pooled quality
control (QC) samples were prepared by transferring a small aliquot of each sample into
a vial (separate QC samples for intracellular and medium metabolites). Selected amino
acids were analyzed with the same method as in targeted UHPLC-MS/MS metabolomics
described above but the MRM transitions were modified (Supplementary Table S7),
allowing analysis of the isotopomers of the selected amino acids. The data was
processed and integrated with MRMPROBS_v.3.66 software (60). Phosphoserine and
glycine were processed and integrated with Agilent Qualitative Navigator B.08.00
software because of their lower abundance. Isotope natural abundances were corrected
with IsoCor 2.2.2 software (61) as earlier reported (62).

Chemical inhibitors

For chemical inhibitor experiments, two different PHGDH inhibitors, NCT-503 (Cat# HY-101966) and CBR-5884 (Cat# HY-100012), were obtained from MedChemExpress. Inhibitor stock solutions were prepared in DMSO, and DMSO was also used as vehicle control. The inhibitors were used in concentrations of 25 or 50 µM, as previously used for different cell types (27, 63–65). Further details of inhibitor experiments are described in the next two sections.

Cell proliferation analysis with EdU labeling

Effects of chemical inhibitors (see the section above) on cell division capacity were evaluated using a Click-iT EdU imaging kit (Cat# C10338, Invitrogen). HUVEC cells were plated onto a 96-well plate in Endothelial cell GM and let to grow in subconfluent culture for 24 h. Exposures were added to the wells, four parallel wells for each condition, and eight parallel wells for the control condition (DMSO vehicle). After 20 h incubation of exposures, 5-ethynyl-20-deoxyuridine (EdU) was added for 4 h. Cells were fixed with 4% PFA for 15 min, washed with 3% bovine serum albumin (BSA) in PBS, and permeabilized with 0.5% Triton X-100 in PBS for 20 min RT, and the Click reaction was done according to the manufacturer's instructions. Before imaging, DAPI (0.5 µg/ml) was added to visualize all nuclei. Cells were imaged using a Nikon A1R confocal microscope. The microscope was automated to find optimal focus in each well, image every well, and take 2-9 views per well so that at least 300 nuclei were imaged. Detected nuclei (all nuclei + EdU-stained nuclei) were analyzed using an in-house built Fiji script (https://github.com/seiryoku-zenyo/simple_ratio_DAPIvsPositive-nuclei/tree/main). Shortly, image data was segmented using a deep learning algorithm, separately for DAPI-positive nuclei and EdU-positive nuclei. Next, the ratio between DAPI and EdU nuclei was counted. The data is presented as a fraction (%) of EdU-positive nuclei out of total (DAPI) nuclei.

Click-iT EdU imaging kit (Cat# C10339, Invitrogen) was also used to determine the effect of PHGDH overexpression on HUVEC proliferation. Cells were cultured on 13 mm diameter coverslips (10 000 cell/cm^2) coated with 0.2% gelatin for 24 h, after which cells were incubated with EdU labeling solution for 5 h. Cells were fixed, washed, and permeabilized as above. Click reaction buffer was prepared as described by the manufacturer, added to the cells, and incubated for 30 min protected from light. DAPI (0.5 µg/ml) was used to detect all the nuclei. Cells were imaged with a confocal microscope (Zeiss LSM700) and the number of stained nuclei was calculated manually with Fiji. The data is presented as a fraction (%) of EdU-positive nuclei out of total (DAPI) nuclei.

Effects of mechanical loading on HUVEC proliferation were assessed both by microscopy and flow cytometry. For both analyses, the cells were mechanically loaded as described earlier and incubated in EdU labeling solution during the mechanical loading (5h), plus an additional 2h after cessation of the loading period. For the microscopy analysis, the cells were treated and imaged as above, directly on the PDMS

chambers. Quantification of EdU-positive nuclei was performed with the aforementioned in-house-built Fiji script. The Click-iT™ EdU Alexa Fluor™ 647 Flow Cytometry Assay Kit (Cat# C10424, Invitrogen) was used for the flow-cytometry analysis. HUVECs were detached after the EdU incubation, and two PDMS chambers were pooled as one sample to ensure a sufficient number of cells for the flow cytometer. Fixing, washing, and staining of the cells were performed following the kit instructions. EdU analysis was performed with CytoFLEX S Flow Cytometer using red 638 nm laser with 712/25 BandPass filter and Cytexpert software 2.4.0.28. Quality control was performed using CytoFLEX Ready-To-Use Daily QC Fluorospheres, and an unstained HUVEC pool was used as a compensation control. 10000 datapoints were collected. Cell debris and determination of singlets with SSC and FSC were performed. Analysis of EdU$^+$ and EdU$^-$ signals was done utilizing histograms.

Carbon-14 labeling

For evaluation of relative protein synthesis and glucose-incorporation the following radioisotope tracers were used: ^{14}C-[1]-L-phenylalanine (0.1 mCi/mL, Cat# NEC859S050UC, Revvity); ^{14}C-[U]-L-valine (271.0 µCi/mmol, Cat# NEC291EU050UC, PerkinElmer); and ^{14}C-[U]-glucose (0.2 mCi/mL, Cat# NEC042V250UC, PerkinElmer). Only one tracer per experiment was used, and the dilution of the tracers was 1 µL/mL (vol/vol) in the culture media. Two milliliters of culture media were used per well of a 6-well plate, and one milliliter per stretching chamber

HUVECs for PHGDH inhibitor experiments were seeded onto 0.2% gelatin-coated 6-well plates at a density of 12,500 cells/cm^2. The cell population was expanded for 2 days, and the inhibition experiments were initiated by changing the media to experimental media containing 25 µM of inhibitor and a radioisotope tracer. After incubating cells for 48 h with inhibitor and tracer, proteins, RNA, and lipids were extracted, and liquid scintillation counting was conducted as previously (27). When protein synthesis was measured during and after mechanical loading, tracer was added to the medium just before the initiation of the 5 h mechanical loading. After the loading, HUVECs were incubated for an additional 20 h in the medium, and then proteins were extracted.

PHGDH overexpression

The PHGDH overexpression construct pLJM5-WT PHGDH was a gift from David Sabatini (Addgene plasmid #83901; http://n2t.net/addgene:83901; RRID: Addgene_83901) (66). Lentivirus particles were prepared and distributed by Biomedicum Virus Core facility supported by HiLIFE and the Faculty of Medicine, University of Helsinki, and Biocenter Finland. The control construct, clone SHC002 (Scrambled shRNA in pLKO.1) was from Sigma Aldrich MISSION shRNA library, distributed by Genome Biology Unit core facility supported by HiLIFE and the Faculty of Medicine, University of Helsinki, and Biocenter Finland. The Biomedicum Virus Core facility also determined the p24 capsid protein concentration with an enzyme-linked immunosorbent assay (ELISA). p24 concentrations were used to calculate the number of virus particles needed for transduction (1 pg of p24 = 100 transducing units).

HUVECs were cultured in T75 flasks. After reaching 50% confluence, 10 ml of medium
containing lentiviruses (MOI = 50) and polybrene (5 μg/ml), was added to the cells.
Polybrene (Cat# TR-1003, Sigma-Aldrich) was used to improve infection efficiency.
HUVECs were incubated with lentiviruses for 24 hours. After transduction, the cells were
resuspended in a fresh endothelial cell medium, cultured overnight, and used for further
experiments. HUVECs expressing scrambled shRNA and HUVECs without virus
treatment were used as controls for the PHGDH overexpression construct.

Protein extraction and Western blotting

HUVECs expressing either the PHGDH overexpression system or shControl gene were
cultured in 6-well plates (5000 cells/cm^2) in endothelial cell medium supplemented with
1 μl/ml (vol/vol) ^{14}C-[1]-L-phenylalanine (see *Carbon-14 labeling* section). After 48 hours,
cells were homogenized in lysis buffer containing 20 mM HEPES (pH 7.4), 1 mM EDTA, 5
mM EGTA, 10 mM MgCl$_2$, 100 mM β-glycerophosphate, 1 mM Na$_3$VO$_4$, 1 mM DTT, 1%
TritonX-100, and supplemented with protease and phosphatase inhibitors (Cat# 78442,
Thermo Fisher Scientific). Cell lysates were centrifuged for 5 min at 13,000 x g, 4 ℃, and
the supernatant was collected. Separate aliquots of the supernatants were used for
western blotting and analysis of ^{14}C-[1]-L-phenylalanine incorporation into proteins with
liquid scintillation counting (see *Carbon-14 labeling* -section). Western blot analysis of
PHGDH was conducted by using low-fluorescence PVDF membranes and anti-PHGDH
antibody (1:5000 dilution, Cat# HPA021241, Sigma-Aldrich) as previously [27] with
GAPDH (1:10 000 dilution, Cat# ab9485, Abcam) as a loading control.

Statistical testing in non-*omics* analyses

Statistical tests were conducted using GraphPad Prism 10.3.1 (Dotmatics). The
Student's t-test was used when two groups were compared, whereas one-way ANOVA
with Fisher's least significant difference test was used when a control group was
compared to various treatment groups. The main effects and interaction of two variables
were examined using two-way ANOVA. Group difference was considered statistically
significant with $p ≤ 0.05$.

Acknowledgments

758 We thank Associate Professor Thomas M. O'Connell for valuable advice regarding ^{13}C
759 isotopic tracing and for providing the R script for ranking the significantly altered genes.
760 We also thank all the laboratory technicians who helped us during this study. Figures 1A
761 and 3A were created with the licensed version of https://www.biorender.com/.

CRediT authorship contribution statement

763 **Sakari Mäntyselkä:** Conceptualization, Methodology, Validation, Formal analysis,
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Funding

This work was supported by the Research Council of Finland (R.K., Grant number: 348636), the Sigrid Jusélius Foundation (R.K.), the Finnish Foundation for Cardiovascular Research (R.K., L. U-K.), the Emil Aaltonen Foundation (S.M., E.N.), and the Finnish Cultural Foundation (K.K. and J.J.H).

Disclosure and competing interest statement

The authors declare that they have no known competing financial interests or personal relationships that could have appeared to influence the work reported in this paper.

Supplementary Data

A PDF file containing Supplementary Figures 1-7.

Excel spreadsheets:

- Table S1: List of DEGs
- Table S2: Overlapping DEGs
- Table S3: GSEA results
- Table S4: Overlapping GSEA pathways
- Table S5: Metabolomics data
- Table S6: ^{13}C-(U)-glucose labeling data
- Table S7: MRM-transitions used in ^{13}C-(U)-glucose tracing

Data availability

Filtered non-aligned RNA sequencing data and non-normalized quantified reads are available in the Gene Expression Omnibus (GEO) data repository (accession number:

GSE288064). Processed transcriptomics data are shared as Supplementary Excel spreadsheets. Numerical metabolomics and isotopic tracing data are shared as Supplementary Excel spreadsheets.

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