

Towards a personalised approach in exercise-based cardiovascular rehabilitation: How can translational research help? A 'call to action' from the Section on Secondary Prevention and Cardiac Rehabilitation of the European Association of Preventive Cardiology

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1 **Towards a personalized approach in exercise-based cardiovascular rehabilitation: How**
2 **can translational research help?**

3 **A 'call to action' from the Section on Secondary Prevention and Cardiac Rehabilitation of the European**
4 **Association of Preventive Cardiology**

5 Andreas B. Gevaert^{1,2,3}, Volker Adams⁴, Martin Bahls^{5,6}, T. Scott Bowen⁷, Veronique Cornelissen⁸, Marcus
6 Dörr^{5,6}, Dominique Hansen^{3,9}, Harel M.C. Kemps¹⁰, Paul Leeson¹¹, Emeline M. Van Craenenbroeck^{1,2}, Nicolle
7 Kränkel^{12,13#}

8
9 # Corresponding author

10 ¹ Research Group Cardiovascular Diseases, GENCOR Department, University of Antwerp, Antwerp,
11 Belgium;

12 ² Department of Cardiology, Antwerp University Hospital (UZA), Edegem, Belgium;

13 ³ Heart Centre Hasselt, Jessa Hospital, Hasselt, Belgium;

14 ⁴ Department of Molecular and Experimental Cardiology, TU Dresden, Heart Centre Dresden, Dresden,
15 Germany;

16 ⁵ University of Greifswald, Department of Internal Medicine B, Greifswald, Germany;

17 ⁶ German Centre for Cardiovascular Research (DZHK), partner site Greifswald, Greifswald, Germany;

18 ⁷ School of Biomedical Sciences, University of Leeds, Leeds, United Kingdom;

19 ⁸ Department of Rehabilitation Sciences, KULeuven, Leuven, Belgium;

20 ⁹ Hasselt University, Faculty of Rehabilitation Sciences, REVAL, BIOMED, Diepenbeek, Belgium;

21 ¹⁰ Máxima Medical Centre, Centre FLOW, Eindhoven, The Netherlands;

22 ¹¹ Oxford Cardiovascular Clinical Research Facility, Radcliffe Department of Medicine, University of Oxford,
23 Oxford, United Kingdom;

24 ¹² Department of Cardiology, Charité Universitätsmedizin, Campus Benjamin Franklin, Berlin, Germany;

25 ¹³ German Centre for Cardiovascular Research (DZHK), partner site Berlin, Berlin, Germany.

26
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29
30 **# correspondence to:** Nicolle Kränkel, PD Dr. rer. nat.
31 Charité – Universitätsmedizin Berlin
32 Campus Benjamin Franklin, Dept. of Cardiology
33 Hindenburgdamm 30, 12203 Berlin, Germany
34 fax: +49-(0)30-450-513999
35 phone: +49-(0)30-450-522246
36 email: nicolle.kraenkel@charite.de
37

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1 **Abstract**

2 The benefit of regular physical activity and exercise training for the prevention of cardiovascular and
3 metabolic diseases is undisputed. Many molecular mechanisms mediating exercise effects have been
4 deciphered. Personalized exercise prescription can help patients in achieving their individual greatest benefit
5 from an exercise-based ~~cardiac~~cardiovascular rehabilitation programme. Yet, we still struggle to provide truly
6 personalized exercise prescriptions to our patients.

7 In this position paper, we address novel basic and translational research concepts that can help us understand
8 the principles underlying the inter-individual differences in the response to exercise, and identify early on who
9 would most likely benefit from which exercise intervention. This includes hereditary, non-hereditary and sex-
10 specific concepts. Recent insights have helped us to take on a more holistic view, integrating exercise-
11 mediated molecular mechanisms with those influenced by metabolism and immunity. Unfortunately, while
12 the outline is recognizable, many details are still lacking to turn the understanding of a concept into a
13 roadmap ready to be used in clinical routine. This position paper therefore also investigates perspectives on
14 how the advent of 'big data' and the use of animal models could help unravel inter-individual responses to
15 exercise parameters and thus influence hypothesis building for translational research in exercise-based
16 ~~cardiac~~cardiovascular rehabilitation.

17

18 Abstract word count: 198

19 Keywords: ~~cardiac~~cardiovascular rehabilitation, exercise, personalized medicine,
20 responders/non-responders, immune system, machine learning, big data, animal
21 models

1 **1. Introduction**

2 Epidemiological and interventional studies have demonstrated a benefit of regular physical activity and
3 exercise for the prevention of cardiovascular and metabolic diseases.¹⁻⁶ Exercise acts in a pleiotropic manner,
4 addressing cardiac contractile and diastolic properties, muscle anabolic and catabolic pathways, substrate
5 metabolism, and regulatory processes governing tissue perfusion and energy storage.^{7,8}
6 While physiological research of the past decades allowed us to understand these principal interactions, crucial
7 questions remain on how to effectively implement exercise interventions in clinical therapy. Access and
8 compliance to cardiovascular rehabilitation (CR) programmes remains a critical factor in the success of an
9 exercise intervention, which requires a highly motivated multi-disciplinary team.⁹ But basic and translational
10 research can also help, addressing questions regarding the personalization of exercise prescription, in order to
11 improve efficacy of exercise interventions throughout the cardio/vascular/metabolic continuum. Why do
12 some patients not respond to exercise-based cardiac rehabilitation (CR), and how can we identify them early
13 on? What drives the difference in response to CR in men and women? How is the response to exercise
14 influenced by metabolism, immunity and their interaction?
15 In addition to that, research methodology is rapidly advancing, bringing different views on translation of
16 biochemical findings into the clinics. How will the advent of 'big data' influence hypothesis building for
17 translational research in CR? What is the sense and nonsense of using animal models in modern CR research?
18 In this position paper, we aim to address these future challenges for basic and translational research in
19 exercise-based CR. We critically review recent studies dealing with the most important yet unanswered
20 questions in the field, both in preclinical and clinical research. Finally, we pinpoint gaps in current evidence
21 that deserve intensified attention in future research.

22
23 **2. Future targets and open questions in translational ~~cardiac~~cardiovascular rehabilitation research**

24 While exercise-based intervention programmes are recommended in primary and secondary cardiovascular
25 prevention,^{10,11} it has also become clear that exercise parameters – type, intensity, duration, frequency –
26 differentially address cardio-vascular and metabolic endpoints, and that the quality and quantity of the
27 response may differ significantly between participants.¹⁰ In addition to improving implementation, it is the
28 personalization of exercise interventions that is an important focus of current and future research.
29 Personalization of therapy includes taking account of patient-specific parameters with potential impact on the
30 mechanism of disease and therapy effect, including age, gender and co-morbidities. In addition,
31 personalization also means that target parameters need to be chosen according to the clinical need of the
32 patient, based on their underlying morbidities and risk profile.

33 **2.1 What factors contribute to the large variability in individual response to cardiovascular rehabilitation? Why**
34 **do some patients not respond?**

35 The improvement in maximal aerobic capacity (peak oxygen uptake, VO₂peak) following exercise-based CR is
36 related to survival in a wide range of cardiovascular diseases, independent of other important risk factors.¹²⁻¹⁴
37 Even small increments in VO₂peak result in substantially lower risk for all-cause and cause-specific mortality.³
38 Although trials that investigated the effects of exercise-based CR on exercise capacity have consistently shown
39 favourable and clinically significant changes,^{15,16} a large variability is seen in the individual training response
40 (relative change in VO₂peak following training, ΔVO₂peak). This variability exists both in healthy subjects and in
41 patients with established cardiovascular disease, when exposed to similar exercise programs.^{14,17,18} Recent
42 studies have shown that up to 33% of patients fail to demonstrate a meaningful increase in VO₂peak in
43 response to CR, despite adequate compliance to training. These 'non-responders' show a decrease in
44 VO₂peak, or an increase within the test-retest variability of VO₂peak measurement (generally accepted to be

±6%).^{18–20} The mechanisms driving this variability in $\Delta\text{VO}_2\text{peak}$ are not well understood, nor do we have good predictors for the response to exercise intervention. Possible contributing factors are summarized in **Figure 1**.

~~We introduce some of the most important contributors below. Interested readers can further explore the main forms of genetic, environmental, and training-related factors that affect this response.~~

Among the factors influencing the individual response to CR, exercise parameters have been studied intensely recently. One way of addressing this line of research is by performing 'secondary' responder analyses on existing datasets of original publications. This can be accomplished by quantifying the number of non-responders and responders to different types of exercise interventions. For example, Williams et al. combined data from different laboratories that had compared training volumes ranging between high and moderate intensities, in populations of both healthy subjects and patients with established cardiovascular disease.²⁴ When exercise was performed with great amounts and high intensities, the likelihood of subjects increasing their exercise capacity was significantly greater. Similarly, Montero et al. showed that healthy non-responders to an exercise training intervention did increase their VO_2peak when subjected to greater training volumes.²⁵ Yet, the evidence regarding the additional beneficial effects of higher exercise intensities is still conflicting.²⁶ Total energy expenditure may be more relevant for. One could argue that the main determinant for improvements in exercise capacity is the total energy expenditure rather than the exercise intensity in these subjects. More comparative exercise intervention studies are needed to determine the inter-individual variability in exercise capacity caused by different variables types of exercise programs. (Figure 3).

It still remains to be elucidated which phenotypic and genotypic characteristics predict the response of a patient to these specific exercise interventions.²³ Previous studies already suggested that in addition to exercise training characteristics (e.g. intensity, volume, type), common personal characteristics like age, sex, body mass index, and baseline physical fitness predict between 15–21% of variability in $\Delta\text{VO}_2\text{peak}$.^{15,17,19,24} Moreover, an additional physiological factor that may influence $\Delta\text{VO}_2\text{peak}$ in patients with chronic heart failure (HF) is the circulatory response to acute exercise.^{27,28} Considering the relatively low predictability of these factors, other more important factors that affect $\Delta\text{VO}_2\text{peak}$ likely still need to be discovered.

Heritability explains more than 50% of the inter-individual differences in cross-sectionally measured VO_2peak .^{29,30} In addition, the Heritage Family study demonstrated that the change in VO_2peak to exercise training intervention is also largely (47%) determined by heritable factors (i.e. genetic, epigenetic or familial environmental factors).³¹ Heritability of training-induced changes in haemodynamic response and skeletal muscle characteristics are also relatively high.^{32,33} Most importantly, the heritability of $\Delta\text{VO}_2\text{peak}$ was independent of baseline VO_2peak .³⁴ This implies that even subjects with a low aerobic capacity may still substantially benefit from exercise training during CR.

Previous research has used several approaches. Single gene diagnostics can help to improve our understanding of the genetics underlying the variability in VO_2peak and $\Delta\text{VO}_2\text{peak}$. In the late 1990s Montgomery et al. assessed left ventricular wall thickness as a proxy for the response to exercise intervention in 140 army recruits.³⁵ Subjects with a specific mutation in the gene for the angiotensin converting enzyme showed no improvement in left ventricular mass, while in subjects without this mutation left ventricular mass increased by 22–26%. After this publication, 'The human gene map for performance and health-related fitness phenotypes' was has identified created.³⁶ Even though more than 200 autosomal gene variants and quantitative trait loci have been identified from this study, the authors noted that as data was mainly derived from underpowered sample sizes, therefore this study did not provide compelling evidence that DNA sequence variants in a given gene are associated with human variation in fitness and performance

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1 traits.³⁵ Interaction between gene variants and disease modifying factors add to the complexity. For example,
2 a single nucleotide polymorphism (SNP) in the *FTO* gene is associated with higher risk for adiposity, but this
3 interaction term was weaker in physically active people.³⁶

4
5 A means to overcome the focus on a single gene or locus could be transcriptome wide RNA expression
6 profiling studies. Timmons et al. identified 11 SNPs in skeletal muscle, which were responsible for nearly 50%
7 of the heritability of $\Delta\text{VO}_2\text{peak}$ in healthy subjects.³⁷ Genome-wide association studies could also provide
8 unbiased insight into the genetics underlying baseline VO_2peak as well as $\Delta\text{VO}_2\text{peak}$. Bouchard et al.
9 discovered a total of 39 SNPs significantly associated with $\Delta\text{VO}_2\text{peak}$.³⁸ Unfortunately, there was no overlap
10 between the genes identified by Timmons et al. and those reported by Bouchard et al.³⁹ Another large
11 genome-wide association study compared SNPs in 1520 elite athletes with SNPs in 2760 non-athletes, and
12 identified only a single SNP (in the *GALNTL6* gene) that was more common in athletes.⁴⁰ Hence, while previous
13 studies have started to use hypothesis-free methods to improve our understanding of the genetics underlying
14 VO_2peak and $\Delta\text{VO}_2\text{peak}$, there is still a long way to go.

15
16 Post-translational Epigenetic modifications regulation – may also influence protein function. This includes DNA
17 methylation, histone modification, and post-translational modifications by non-coding RNAs, and each of
18 these mechanisms has been described to contribute to the response to exercise training. Indeed, several non-
19 coding RNA species have been linked to VO_2peak or $\Delta\text{VO}_2\text{peak}$.

20 Both acute bouts of exercise and repeated training influence promoter DNA methylation.^{41–43} Acute exercise-
21 induced expression of key signalling pathways, including AMPK/PGC-1 α , was paired with a hypomethylation of
22 the respective promoter sequence.⁴² Importantly, the magnitude of the effect on DNA methylation was
23 dependent on exercise dose, suggesting a role of DNA methylation in the individual response to training.⁴²
24 Deacetylation of histones and – but also other proteins, by sirtuins, is known for a while to mediate adaptation
25 to repeated exercise.⁴⁴ Exercise induced histone modifications, including acetylation and methylation, have
26 also been reported.^{44–46}

27 failure depending on which metabolic pathway is switched on when the heart is put under stress.⁴⁵ PMID: 29327474 In
28 addition, histone deacetylase 3 (HDAC3) plays a major role in skeletal muscle by regulating fuel metabolism.⁴⁶ PMID: 26438923 These
29 findings are especially interesting with regards to the insulin resistance in patients with metabolic disease syndrome.
30 Whether or not pharmaceutical interventions targeting HDAC3 histone deacetylases add an additive effect to exercise
31 based CR alone, remains to be determined by future research.

32
33 Finally in healthy subjects, microRNAs are released into the circulation even after acute exercise, and exercise
34 training induces long-term changes in their expression.^{47,48} In a rat model of HF, Souza et al. identified a set of
35 14 cardiac microRNAs of which expression was influenced by exercise training.⁴⁹ Other studies have identified
36 additional exercise-responsive microRNAs in animal models of different cardiovascular diseases.⁴⁸ To date,
37 only two small studies have assessed the effect of exercise training on microRNA expression in human
38 patients with established cardiovascular disease.^{50,51} Taurino et al. showed that *miR-92a* and *miR-92b* were
39 upregulated after exercise-based CR in patients with coronary artery disease, coinciding with a
40 downregulation of their gene targets.⁵⁰ Xu et al. identified 3 microRNA dysregulated by acute exercise in HF
41 patients, but a clear correlation with VO_2peak was not found.⁵¹

42 None of these epigenetic mechanisms has yet been linked to $\Delta\text{VO}_2\text{peak}$. More Exercise epigenetics is a highly active
43 research area, and more extensive studies, including larger numbers of patients, and screening for more miRNA, are still needed before
44 reliable conclusions can be drawn.

1 While for most studies, improvement of VO_2peak is the main target parameter of an exercise intervention. Yet,
2 depending on the population and the parameters selected as a target, the clinical benefits of exercise training are not limited to VO_2peak .
3 parameters such as VO_2peak . This is not necessarily always VO_2peak , but can be improved submaximal exercise
4 parameters, increased cardiac function, better glucose handling, reduced inflammation, or improved vascular
5 stiffness might be more relevant too.^{5,52-55} Of note, target parameters of the exercise intervention might even
6 change over time in each patient.

- 7
- 8 ➤ To summarize, the change in VO_2peak to exercise training shows large inter-individual variability.
9 Understanding how such inter-individual differences emerge is important, as a lower response is linked
10 to poorer outcomes.¹²⁻¹⁴ $\Delta\text{VO}_2\text{peak}$ seems to be regulated by the interaction between heritable factors
11 and lifestyle – including exercise parameters, SNPs, and non-coding RNAs – but individual targets have
12 yet to be confirmed. We need controlled randomized studies using multi-omics techniques
13 (transcriptomics, genomics, proteomics and metabolomics) to identify potential pathways in a
14 ‘systems biology’ approach. The complex interaction between lifestyle and heritable factors likely
15 explains a large part of the individual response to exercise training, and future studies should aim to
16 improve our understanding of this interaction.

17 2.2 The potential role of sex differences in response to CR

18 In general, VO_2peak is ~15% lower in women compared to men.⁵⁶ Intriguingly, however, women seem to
19 experience better clinical outcomes following exercise training, despite similar improvements in exercise
20 capacity.^{57,58} While sex-specific effects thus likely play a key role in the clinical benefits associated with
21 exercise interventions, the mechanisms responsible for these benefits remain poorly understood.
22

23 Cardiovascular physiology as well as pathophysiology are markedly different between men and
24 women, as has been reviewed extensively elsewhere. These differences are summarized in Table 1. For more details, see the full text of the review.^{59a}

25
26 Sex-specific hormones may explain part of these differences. In pre vs. postmenopausal women of similar age,
27 blood pressure is lower, and left ventricular end-systolic volume, ejection fraction and filling rate are larger.⁶²
28 The vasodilating properties of oestrogen may play a role.⁶³ Also, RNA sequencing in cardiomyocytes revealed
29 more than 600 genes with sexually dimorphic expression patterns.⁶⁴ This adds to genetic differences due to
30 male specific Y-chromosomal gene expression and differences in epigenetics (histone and DNA modifications,
31 non-coding RNA expression).⁶⁰
32 Thus, in addition to the obvious endocrine differences between men and women, a variety of anatomical,
33 genetic and molecular differences exists within the heart. These may influence not just cardiovascular
34 disease progression, but also affect secondary prevention strategies.⁶¹

35
36 While central hemodynamic differences likely explain some of the sex-specific effects in response to CR,⁶¹
37 other factors are also involved. It is well established that cardiovascular disorders induce secondary
38 impairments to the periphery, including endothelial and skeletal muscle dysfunction, which are closely linked
39 to symptoms of exercise intolerance and prognosis.⁶⁵ Surprisingly, it is still largely unclear how sex modulates
40 the crosstalk of mechanisms governing the loss of endothelial, skeletal and cardiac function. A few studies
41 have revealed that in patients with HF, mitochondrial enzymes in skeletal muscle show either no major
42 changes or more pronounced deficits in men compared to women, with a greater shift towards glycolytic
43 enzymes and type IIX fatigable fibres in men.^{66,67} In response to an aerobic endurance training intervention,
44 evidence has revealed minor differences in terms of skeletal muscle biochemistry, with reports suggesting
45 men with HF can increase the content of the slow myosin heavy chain isoform towards similar levels to that

1 observed at baseline in women.⁶⁸ Thus, women may experience a greater preservation of muscle oxidative
2 function compared to men with HF, which could help to explain why women demonstrate greater clinical
3 benefits after CR.⁵⁷ The mechanisms underpinning the sex-specific differences in muscle physiology and
4 effects of exercise intervention remain unclear. Hormonal effects of oestrogen regulation on mitochondrial
5 dynamics and/or a preferential shift towards fatty acid oxidation in women may play a role,^{69,70} but more
6 extensive measures of muscle function and physiology and higher sample sizes are still required to confirm
7 this.

8
9 In addition to skeletal muscle alterations, endothelial dysfunction also develops in HF patients, both in men
10 and women.⁷¹ Yet, little data is available to clearly demonstrate whether any sex-specific alterations are
11 present following CR in patients. Recent evidence from animal models of HF have shown that high-intensity
12 interval training can attenuate endothelial dysfunction in both female and male rats, which seems to act via
13 mechanisms specifically lowering oxidative stress in males and increasing endothelial nitric oxide synthase
14 expression in females.^{72,73} Whether these molecular benefits are paralleled in male and female patients with
15 HF remains unclear. Furthermore, sex-specific substrate utilisation could play a key role in the exercise
16 response in women and may fill the above-mentioned gap in the literature with regards to the effectiveness
17 of exercise-based CR. One example is that women rely on carbohydrates to a lesser extent but have a higher
18 content of intramyocellular lipids.⁷⁴

19
20 While CR programs clearly reduce the risk of all-cause and cardiac-related mortality and improve quality of
21 life, directly extrapolating these findings from men to women remains fraught with complexities since women
22 have consistently been under-represented in previous trials.⁷⁵ In large meta-analyses and randomized
23 controlled trials, the amount of women recruited was 11-28%.⁶¹ Given that women are also ~40 % less likely
24 to enrol in CR and have a significantly lower adherence to the interventions compared to men,^{76,77} the need
25 to better understand sex-specific mechanisms in response to exercise training will first require rapid
26 improvement in CR recruitment and adherence of women. Identification of sex-specific targets is likely to
27 substantially improve outcomes following CR programmes by optimising training regimes.
28 Nonetheless, women seem to benefit at least as much from exercise-based CR as men.^{57,78,79} The most recent
29 Cochrane reviews which assessed the benefits of exercise-based CR concluded that exercise improves
30 cardiovascular mortality and hospitalization (in patients with coronary artery disease) and improves health-
31 related quality of life (in patients with coronary artery disease or HF).^{80,81} The authors also clearly state that
32 evidence for benefits of exercise-based CR in women is currently insufficient. Given the above mentioned
33 physiological and pathophysiological differences between men and women, we cannot assume that exercise
34 regimes which worked for men will also be effective for women.

35
36 ➤ *To summarize, important differences exist in the response to CR in men and women. Besides obvious*
37 *differences in cardiovascular and skeletal muscle structure, function and physiology, the underlying*
38 *hormonal and molecular mechanisms are still understudied. Identification of sex-specific targets might*
39 *further improve outcomes after CR. Further, in order to put the physiological differences between*
40 *men and women into a larger perspective, novel 'omics' techniques, omics approaches which enable a systems biology*
41 *approach, should be used to determine which differences contribute to the response to exercise based CR.*

42 **2.3 Immune-metabolism interactions and inflammation**

43 Both enhanced activation and impaired resolution of inflammation are major underlying principles of
44 cardiovascular and metabolic pathologies.⁸² Regular exercise training has been shown to lower systemic and
45 vascular inflammatory load within a few weeks of intervention.⁵⁵ This has been partly attributed to active

1 secretion of anti-inflammatory myokines from skeletal muscle.⁸³ While biochemical interactions of some
2 myokines have been deciphered, it remains a major task to chart the network of biochemical interactions
3 between energy demand by skeletal muscle contractile activity (affected by exercise parameters, such as
4 duration, type and frequency) and the fine-tuning of inflammatory mechanisms. The recent years have
5 brought a refinement in our understanding of inflammation in atherosclerosis, including the appreciation of
6 resolution of inflammation as an active process, distinct from inhibition of inflammation, as well as the tight
7 interactions between immune cell activation and their energy metabolism. ~~Only a fraction of these newly~~
8 ~~understood mechanisms has been exploited in the context of the cardiovascular effects of regular exercise~~
9 ~~training.~~ Those initial in vitro data have not yet been translated into ~~therapy~~ therapeutic strategies. Unanswered questions
10 include to which extent immunometabolic observations made in mouse macrophages can be translated to the
11 human and to which extent in vitro differentiated macrophage phenotypes resemble in vivo macrophages,
12 regarding both, immunologic function and energy metabolic profile.

13
14 Resolution of inflammation versus anti-inflammation: The termination of an acute inflammatory response is
15 normally governed by both the decay of pro-inflammatory signals, as well as the active production of pro-
16 resolving factors.⁸⁴ The inability to resolve an ongoing inflammatory process is a hallmark of inflammatory
17 degenerative ~~processes~~ diseases, including atherosclerosis.⁸⁵ On the one hand, innate immune-activating
18 signals - ligands of pattern-recognition receptors, such as modified lipids - do not disappear in atherosclerosis,
19 as it would happen in a 'normal' injury. On the other hand, the production of pro-resolving mediators appears
20 to be dysregulated. Anti-inflammatory therapies have been employed more or less successfully in secondary
21 cardiovascular prevention.^{86,87} However, therapeutic success appears to depend on the inflammatory
22 signalling mechanism targeted, likely interleukin-1 β and interleukin-6 signalling, and may be flawed by
23 increased incidence of lethal infections.^{86,87} In addition, blocking inflammation also appears to block resolving
24 mechanisms, the removal of apoptotic particles and cell debris as well as the induction of regenerative
25 processes.⁸⁵

26 ~~The termination of an acute inflammatory response is normally governed by both the decay of pro-~~
27 ~~inflammatory signals, as well as the active production of pro-resolving factors.~~⁸⁸ A number of studies support
28 the ability of exercise - ranging from a single session of high-intensity interval exercise to a 3-month
29 multicomponent exercise programme - to reduce cellular responsiveness to toll-like receptor -mediated
30 signalling, induced by damage-associated molecular patterns.⁸⁸⁻⁹⁰
31 Dietary interventions targeting synthesis of specialized resolving mediators (SPM) have been tested for some
32 time now and it becomes evident that both, the dosage and the formulation might be relevant to their
33 success in cardiovascular prevention.⁹¹ In contrast, only few studies have systematically addressed the effects
34 of exercise intervention on the release of SPMs - resolvins, lipoxins, protectins and maresins - but the existing
35 literature indicates an increase in SPM release by regular exercise.⁹²⁻⁹⁴ This might be attributed to acute and
36 chronic effects: strain and acute release of pro-inflammatory mediators are associated with SPM release in
37 acute high-intensity exertion, while chronic effects of exercise intervention might be connected to the
38 exercise-mediated shift in macrophage polarization towards the M2-like phenotype.^{92,94,95} M2-like
39 macrophages are better suited to perform efferocytosis than the M1-like phenotype and it is during
40 efferocytosis that SPMs are released.⁹⁶ Thus, we know that regular exercise is associated with a shift towards
41 the more pro-resolving macrophage spectrum, as well as higher levels of pro-resolving mediators, but we do
42 not know which exercise parameters (e.g. intensity, volume, type) could be used to boost this effect, nor
43 whether a combination with dietary approaches to supplement SPMs could potentiate the effects of exercise
44 intervention on cardiovascular inflammation (**Figure 2**).

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1 Energy metabolism and inflammation: From tumour biology, we know that increased glycolysis and
2 glutaminolysis provide energy flexibility to the cell and generate intermediates that feed into anabolic
3 processes - probably the reason why glycolysis is preferred over oxidative phosphorylation by proliferating
4 tumour cells.⁹⁷⁻⁹⁹ In a similar manner, glycolysis is preferred by activated and proliferating myeloid and
5 lymphoid cells¹⁰⁰ and stimulating glycolysis can activate macrophages.¹⁰¹ In addition, M1-type macrophages
6 feature a 'broken' Krebs cycle, with increased output of intermediates that serve as substrates in the synthesis
7 of pro-inflammatory mediators, or are pro-inflammatory mediators in themselves.¹⁰²⁻¹⁰⁴ In contrast,
8 'alternative' M2-like macrophages favour oxidative phosphorylation and fatty acid oxidation.^{105,106} Indeed,
9 oxidative phosphorylation ~~was~~ is a prerequisite of M2-type phenotypic macrophage polarization.¹⁰⁶

10 ~~Interaction between mitochondrial biogenesis and inflammation here in addition to the 'repurposing' of the Krebs cycle to derive inflammatory intermediates, mitochondrial~~
11 ~~mass and density are affected by inflammation. Inflammation induces mitochondrial damage and reduces mitochondrial mass and density, while exercise and~~
12 ~~species - potentially indicative of mitochondrial damage but also increased in inflammation - upregulates anti-inflammatory and mitochondrial~~
13 ~~repair. Biogenesis programmes leading to increased mitochondrial mass/density in inflammation.¹⁰⁷ Similarly, reactive oxygen~~
14 ~~species have been shown to be a crucial signalling mediators in exercise training, including ~~in this paper~~ and exercise training exercise-induced~~
15 ~~also upregulates activation of AMPK/PGC-1 α signalling, inducing general anabolic pathways as well as mitochondrial biogenesis. ~~add references~~.^{108,109} Obviously,~~
16 ~~and essential signalling pathways including the mitogen-activated protein kinase (MAPK) and related kinases (ERK1/2 and JNK2) and nuclear factor- κ B~~
17 ~~and the protein kinase B are employed in inflammation as well as in exercise, with partial overlaps and divergencies. Similar to the severity of~~
18 ~~inflammation, exercise intensity appears to modulate activation of individual MAPK signalling~~
19 ~~pathways.~~^{107,110,111}

21 ~~Interaction between mitochondrial biogenesis and inflammation here in mitochondrial damage versus repurposing of the Krebs cycle, reduced mitochondrial biogenesis~~
22 atherosclerosis, have not been charted in detail for their inflammation-resolving and energy metabolism
23 phenotype yet, nor regarding the effect of exercise in their polarization. Similarly, NK cells and various T
24 lymphocyte populations react to acute and chronic exercise and contribute to both, polarization of innate
25 immune cells and functionality of various tissues and organs, including distinct fat depots (perivascular,
26 subcutaneous, visceral).¹¹²

28 Both the amount and type of energy substrates provided and physical exercise can affect the phenotype of
29 monocytes and macrophages.^{101,113-116} Energy sensors, such as AMP-dependent ~~kinase-kinase~~, can be
30 targeted by both diet and exercise. On the way to personalized lifestyle-based therapies, we need to learn
31 more about the integration of exercise parameters (e.g. type, intensity, frequency, volume) with diet (e.g.
32 macronutrient composition, amount and timing of eating/fasting) and pharmacological means to modulate
33 energy metabolism and (thereby) activation state of inflammatory cells in various tissues.¹¹⁷⁻¹²¹ Of note,
34 activation of the relevant mechanisms might shift between individuals, being influenced by a number of
35 factors such as hormonal status/sex, age, pharmacotherapy and co-morbidities as well as genetic
36 background.¹²²⁻¹²⁴

38 ➤ *To summarize, macrophage phenotype shift, leading to reduced release of pro-inflammatory mediators
39 and an increased release of pro-resolving mediators, might well be a nexus of exercise-mediated anti-
40 inflammatory and metabolic cardio-protective effects. The available seminal data, however, requires a
41 better resolution: continuously improved techniques of single-cell immuno-phenotyping¹²⁵ and assessment
42 of cellular metabolism¹²⁶ allow for the fine-mapping of immune-inflammatory interactions and can be
43 used to develop diagnostic tools, assessing individual response to exercise and personalizing exercise
44 parameters. In addition, better understanding of the cellular and molecular nodes of the immuno-
45 metabolic network might help to optimize exercise parameters on an individual level to improve cardio-*

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1 *vascular and metabolic benefit, potentially in combination with pharmacological and diet-based*
2 *approaches.*

3. Challenges and opportunities in translational CR research methodology

5 The advent of high-throughput molecular techniques, single-cell diagnostics and organs-on-a-chip have
6 opened countless opportunities in exercise research, but some important challenges have surfaced
7 simultaneously.^{127,128} How can we successfully pinpoint important findings within these vast datasets? And if
8 computers can handle increasingly complex tasks, what is the use of animal models in the future?

3.1 Impact of 'big data' and artificial intelligence on translational research in CR

10 As analytical techniques evolve, new challenges arise with regards to handling the enormous amount of data
11 they generate. This is especially true in the area of genomics, epigenomics, proteomics and metabolomics, but
12 also applies to datasets obtained from large clinical trials or registries, and epidemiological research.¹²⁸ These
13 datasets cannot be readily viewed on any computer, which complicates human pattern recognition.
14 Moreover, the analysis of 'big data' requires additional statistical precaution, taking into account the
15 increased 'noise' of high-throughput techniques.¹²⁷ Novel 'data mining' techniques have been developed to
16 derive relationships and statistical inference from these datasets, often relying on some form of artificial
17 intelligence. These techniques, grouped under the term 'machine learning', can be either supervised (the user
18 determines the relation between subjects) such as traditional regression analysis, or unsupervised (the
19 computer determines the relation between subjects), such as clustering analysis.^{129,130}

21 Some of these novel techniques have already been applied to translational exercise research. In 2009, Goud
22 et al. set up a cluster-randomized trial in 21 CR centres, comparing effects of a computerized decision support
23 system to standard care.¹³¹ In centres implementing the decision support system, concordance with CR
24 guideline recommendations were modestly increased, reducing both over- and under-treatment. Further
25 efforts have been made with regard to artificial intelligence-based exercise prescription.¹³²⁻¹³⁶ Most of these
26 studies describe a framework to automate exercise prescription based on patient demographics,
27 comorbidities, test results and reason for referral. Randomized clinical trials evaluating ~~these~~ fully
28 computerized exercise prescription are still lacking.

29 Finally, the vast amount of data obtained from wearable devices opens up possibilities for data-driven
30 personalization strategies. For example, one study succeeded in predicting active energy expenditure (a
31 predictor of $\Delta\text{VO}_2\text{peak}$) from photo-plethysmographic heart rate measurements, even in patients under beta
32 blocker therapy.¹³⁷

34 But many more possibilities of 'big data' and machine learning exist in the field of CR, which we will
35 demonstrate at the hand of two examples from other areas within cardiovascular research: imaging and
36 phenotyping.

37 Imaging is especially suited for the application of machine learning because images contain a rich amount of
38 data both within the image itself and through extraction of quantitative features.¹²⁹ Furthermore, powerful
39 computational approaches to handle image data have undergone extensive development within academic
40 clinical research and non-medical fields such as facial recognition and image searching.¹³⁸ Combined with
41 recent availability of large imaging datasets,¹³⁹ this has meant artificial intelligence approaches to identify
42 images, automatically quantify image features and predict disease from the patterns in the image have
43 developed rapidly within cardiology and radiology.¹²⁹ As a result, automated quantification is now entering
44 clinical use, but broader diagnostic application will require robust clinical validation before adoption.¹⁴⁰ Of

1 particular interest in CR will be to understand whether imaging after ~~cardiac~~cardiovascular events (e.g.
2 echocardiography) contains information of value for prediction of outcome, risk of HF and likelihood of
3 response to exercise interventions.

4
5 Another approach of unsupervised machine learning is to find clusters of similar data items: subjects in the
6 same cluster are similar to each other, and dissimilar to subjects in other clusters. This can aid in discovering
7 subtypes of patients with a certain disease. For example, machine learning has been able to identify clusters
8 of patients with HF based on their baseline characteristics and test results (including cardiopulmonary
9 exercise tests).¹⁴¹⁻¹⁴⁴ Phenotyping through machine learning predicted the prognosis of HF patients, and
10 performed better compared to traditional predictors such as ejection fraction.¹⁴³

11
12 A major concern of artificial intelligence is the 'black box' phenomenon. More complex machine learning
13 processes, such as neural networks, build layer upon layer of automated decisions up to a point where it is
14 impossible to retrace the individual steps.¹⁴⁵ Thus, while some neural networks have been proven to
15 outperform humans (for example in image recognition¹⁴⁶), it is often hard to assess *how* the computer
16 reached its decision or classification. One technique to overcome the 'black box' is to ask the computer to
17 simultaneously create a simpler 'surrogate' model to gain insight in the reasoning process.¹⁴⁷
18 Also, while the decision process can be fully automated and intelligent, large datasets still need to be imputed
19 to train machine learning models. Availability of enough training data is currently still an issue, but the
20 increased promotion of open science and data sharing will hopefully provide an answer to this problem
21 soon.¹⁴⁸ [For example, several platforms have been set up to share anonymized cardiac imaging data with the](#)
22 [goal of promoting its use in machine learning applications.](#)¹⁴⁹

23 Finally, a major challenge will be to convert artificial intelligence-derived predictions and recommendations
24 into effective action. Better phenotyping and improved risk stratification do not automatically lead to improve
25 health. To truly achieve a health care transformation, behavioural changes are needed at both patient and
26 physician level.¹⁵⁰ For example, artificial intelligence may improve exercise prescription, but a patient's health
27 will only improve if his or her physician implements this improved prescription in practice, and he or she
28 adheres to the prescribed training.

29
30 ➤ *To summarize, early applications in CR research and advanced examples from imaging and*
31 *phenotyping studies show that the advent of 'big data' and machine learning will likely change current*
32 *practice. Major challenges include picking up useful signals between increased noise in big datasets,*
33 *the 'black box' phenomenon, and implementing behavioural changes based on computerized*
34 *recommendations. We suggest some ~~future research areas~~approaches in ~~Figure 3. Table 2.~~*

36 3.2 Sense and nonsense of animal models

37 Appropriate animal models are important to unravel the molecular mechanisms for how exercise-based CR
38 mediates its beneficial effects. Small rodents in particular are attractive models for cardiovascular research,
39 possessing unique properties such as easy handling, short gestation time and low costs. Perhaps most
40 important is the availability of transgenic mice and rats, which allow the possibility to study the involvement
41 of specific molecules in transmitting the positive effect of exercise training, which otherwise would not be
42 possible in humans. Nevertheless, a certain scepticism is warranted based on whether animal models
43 appropriately translate to humans, which has resulted (and rightly so) in the value of such research being
44 questioned.¹⁵¹⁻¹⁵³

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Table 2
I think a figure could be used instead of this table. Could authors provide a figure representing the concept reported in table 2
I mean something like a flow chart representing the research pathway, going from basic research to clinical implementation of new concepts.
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1 An ideal disease model should mimic the human condition genetically, experimentally and physiologically.
2 Therefore, using inbred mouse strains may not reflect the response generated in a genetically polymorphic
3 human population, which may be one reason for the failure of many promising preclinical drugs when
4 translated into human clinical trials. In support, a recently published comment stated that >80% of potential
5 therapeutics fail when tested in humans, even after animal studies have provided evidence that the treatment
6 is safe and effective.¹⁵⁴ One future avenue to circumvent such translational problems may reside in the use of
7 humanized models, whereby mice expressing human transgenes or engrafted human cells/tissue are used in
8 preclinical research.¹⁵⁵ Obviously, generating diseased animal models due to genetic defects is much easier
9 than trying to mimic a more complex disease pattern, where several comorbidities contribute to the final
10 clinical phenotype. One contemporary example of such a complex disease is heart failure with preserved
11 ejection fraction (HFpEF). Since the development of HFpEF is driven by several comorbidities, which include
12 hypertension, diabetes, obesity and ageing,¹⁵⁶⁻¹⁵⁸ it remains difficult to define an animal model that
13 appropriately mimics the HFpEF phenotype. As of yet, the animal models used to probe molecular changes
14 occurring in HFpEF and in response to exercise training have been predominantly based on a single risk factor
15 such as aging or hypertension.^{72,159,160} More recently this line of research included a more clinically relevant
16 animal model, in a way that HFpEF develops due to the onset of multiple comorbidities that mirror a
17 metabolic syndrome.^{73,161-163} Another problem with appropriate animal models may be that most models
18 develop over a short time period, whereas in humans sometimes several years or decades pass before a clear
19 phenotype is established.

20
21 Animals used for cardiovascular exercise studies most commonly range from small rodents (e.g. mice, rats) to
22 large animals (e.g. rabbits, canine, goats, sheep, pigs, horses).¹⁶⁴⁻¹⁶⁹ In these animal models exercise can either
23 be voluntary (e.g., animal cage is equipped with a running wheel) or forced (e.g., animal is placed onto a
24 treadmill for a specific period). Many exercise training studies have been employed using a variety of animal
25 models of diseases that include HF,^{161,170,171} diabetes,^{172,173} and neurodegenerative diseases.¹⁷⁴ Beside the
26 classical animal models (mouse and rat) used to analyse the effect of exercise training on molecular and
27 physiological parameters, more recently other species have been used such as drosophila and zebrafish.¹⁷⁵⁻¹⁷⁸
28 Exercise training in drosophila results in improvements of physiological and molecular measures, which
29 include enhanced climbing speed, flight performance, aconitase levels, and cardiac contractility. Clearly, while
30 the main advantage of using flies as an animal model is that you can train several thousand flies
31 simultaneously, the question of whether and to what extent these findings translate to humans looms large.
32 We also have to keep in mind that it is even more difficult in animal models to control for activity levels. In
33 human studies most of the patients recruited into an exercise study exhibit a very low exercise level, which is
34 difficult to control for in animals.

35
36 ➤ *To summarize, the 'sense' to use animal models to investigate the benefits associated with exercise in
37 disease is difficult to refute: animal studies have often provided the initial clues to help elucidate how
38 exercise exerts its benefits for treating disease. However, animal research can also provide much
39 'nonsense' when translated to humans. Future studies should therefore continue focusing on developing
40 more complex and robust animal models of disease that closely reflect the human condition.*

42 **4. Conclusion and Outlook**

43 Exercise-based CR has consistently shown positive effects on the course of cardiovascular disease.
44 However, recent studies showed that there is a large variation in training effects at the individual level, with
45 up to one third of patients failing to demonstrate a significant increase in exercise capacity despite adequate
46 compliance. Therefore, in order to improve the effects of exercise-based CR it is crucial to (1) gain more in-

1 depth knowledge on the determinants and mechanisms governing the response to exercise in the organs -
2 beyond the skeletal muscle, heart and vascular system - and (2) to acknowledge their interaction at a systemic
3 level.

4 Heritable and non-heritable factors each determine approximately 50% of inter-individual heterogeneity in
5 $\Delta\text{VO}_2\text{peak}$. High-throughput technologies in combination with improved bio-informatics and bio-statistical
6 approaches can help identify major regulatory nodes among large datasets that cannot be readily interpreted
7 otherwise.

8 Sex-specific differences in the response to exercise in cardiovascular therapy are severely understudied.
9 Although endocrine, anatomical and molecular differences between men and women are assumed to play a
10 role, the exact mechanisms remain largely unknown. Future research therefore needs to include sufficient
11 numbers of female patients to address these issues.

12 Based on these studies, a concise, easy-to-use panel of markers that could help personalize exercise
13 parameters could be developed. This panel could include regulatory nodes identified in clusters of patients
14 through their classical risk profile, but also inflammatory and metabolic status, and genetic traits identified
15 through advanced bio-statistics. Finally, while animal models have inherent limitations complicating
16 translation to humans, complex and robust animal models closely reflecting human cardiovascular diseases
17 will be needed to test the hypotheses mentioned and to gain further insight in the complex physiology of
18 exercise-based CR.

19 **Abbreviations:** EF = ejection fraction, LV = left ventricular

20 AI = artificial intelligence, CR = cardiac rehabilitation, SNP = single nucleotide polymorphism

21 **Figure Legends**

22 **Figure 1:** Known factors possibly influencing the response to exercise training. These factors are grouped as
23 cardiac, non-cardiac, external and comorbidities. They possibly influence- baseline VO_2peak and/or $\Delta\text{VO}_2\text{peak}$,
24 and are themselves determined by genetic, epigenetic, and environmental factors and drugs, nutrition and
25 sex. CKD = chronic kidney disease, COPD = chronic obstructive pulmonary disease, VO_2peak = peak oxygen
26 uptake, $\Delta\text{VO}_2\text{peak}$ = relative change in VO_2peak following exercise training.

27
28 **Figure 2:** Known and unknown interactions between exercise, nutrition and pro-resolving macrophage
29 polarization and function in cardiovascular disease. CVD = cardiovascular disease, DAMP = damage-associated
30 molecular patterns.

31
32 **Figure 3:** Suggested research areas for application of data mining and machine learning in exercise-based CR.
33 Left column: research questions or clinical needs in the area of exercise-based CR in which data mining and
34 machine learning could play a role. Middle column: suggested data sources for machine learning input. Right
35 column: examples and references. AI = artificial intelligence, CR = cardiovascular rehabilitation, SNP = single
36 nucleotide polymorphism

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9 **Conflicts of Interest**

10 The Authors declare that there is no conflict of interest.
11

12 **Author contributions**

13 All authors contributed to the conception or design of the work, to the acquisition, analysis, or interpretation
14 of data for the work. All authors drafted the manuscript. All authors critically revised the manuscript, gave
15 final approval and agree to be accountable for all aspects of work ensuring integrity and accuracy.
16
17

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