Genetics and the heart rate response to exercise

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Abstract
The acute heart rate response to exercise, i.e., heart rate increase during and heart rate recovery after exercise, has often been associated with all-cause and cardiovascular mortality. The long-term response of heart rate to exercise results in favourable changes in chronotropic function, including decreased resting and submaximal heart rate as well as increased heart rate recovery. Both the acute and long-term heart rate response to exercise have been shown to be heritable. Advances in genetic analysis enable researchers to investigate this hereditary component to gain insights in possible molecular mechanisms underlying interindividual differences in the heart rate response to exercise. In this review, we comprehensively searched candidate gene, linkage, and genome-wide association studies that investigated the heart rate response to exercise. A total of ten genes were associated with the acute heart rate response to exercise in candidate gene studies. Only one gene (CHRM2), related to heart rate recovery, was replicated in recent genome-wide association studies (GWASs). Additional 17 candidate causal genes were identified for heart rate increase and 26 for heart rate recovery in these GWASs. Nine of these genes were associated with both acute increase and recovery of the heart rate during exercise. These genes can be broadly categorized into four categories: (1) development of the nervous system (CCDC141, PAX2, SOX5, and CAV2); (2) prolongation of neuronal life span (SYT10); (3) cardiac development (RNF220 and MCTP2); (4) cardiac rhythm (SCN10A and RGS6). Additional 10 genes were linked to long-term modification of the heart rate response to exercise, nine with heart rate increase and one with heart rate recovery. Follow-up will be essential to get functional insights in how candidate causal genes affect the heart rate response to exercise. Future work will be required to translate these findings to preventive and therapeutic applications.

Keywords Heart rate increase · Heart rate recovery · Exercise · Genetics

Introduction
The regulation of resting heart rate is complex; autonomic tone, central and peripheral reflexes, hormonal influences, and factors intrinsic to the heart are all important determinants [1, 2]. Despite recent developments in the understanding of the complex interplay of the plethora of biological mechanisms influencing resting heart rate [3], our understanding is still incomplete.

The acute heart rate response to exercise, heart rate increase during and heart rate recovery after exercise, offers unique insights into cardiac physiology compared to heart rate in rest and can therefore be exploited to obtain additional information on cardiac function [4]. Impaired increase of heart rate during exercise (chronotropic incompetence) and an attenuated heart rate recovery have been associated with all-cause mortality and sudden cardiac death in healthy individuals [5–7] and in those with cardiac disease, including individuals with heart failure [8] and coronary artery

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disease [9]. Regular endurance exercise training has been proven to shift the cardiac autonomic balance towards vagal dominance [10]. The long-term response of heart rate to exercise results in favourable changes in chronotropic function, including decreased resting and submaximal heart rate as well as increased heart rate recovery [11].

Both the acute and long-term responses of heart rate to exercise have been shown to have a large heritable component [12–17]. Development in the understanding of the human genome and genetic analysis enables researchers to investigate the possible molecular mechanisms underlying interindividual differences in the acute and long-term heart rate response to exercise [18]. In this review, we summarize the current knowledge of the acute and long-term heart rate response to exercise, with a focus on the genetic contribution. In addition, we identify gaps in our knowledge and discuss possible future directions that might be of interest to enhance the understanding of the heart rate response to exercise and consider its potential clinical applications.

### Acute response

#### Heart rate increase

In general, the regulation of the circulatory system during exercise involves several adaptations. These adaptations include dilatation of resistance vessels in the active muscles, a decrease in vagal outflow to the heart, followed by an increase of sympathetic outflow. If exercise is intense, the cholinergic fibers to the adrenal medulla are also activated, resulting in release of epinephrine into the circulation [19]. Under normal physiological conditions, this results in increased venous return, contractility, and heart rate [20]. In turn, ejection fraction increases due to a greater ejection of blood at the end of systole and increased diastolic filling of the ventricles as the duration of the systole decreases with increased heart rate [20].

The increase of heart rate during exercise is for a major part attributable to the decrease in vagal tone followed by an increase in sympathetic outflow and an increase in levels of circulating catecholamines [19]. It has been shown that a substantial component of interindividual differences in the heart rate increase during exercise is genetically determined, with heritability estimates ranging from 0.17 to 0.32 (Table 1) [12, 14, 15]. This suggests that genetic analyses may identify novel biological mechanisms involved in the regulation of heart rate response to exercise.

Several studies have focussed on identifying genetic determinants that explain interindividual differences in heart rate increase during exercise. Genes investigated in these studies are summarized in Table 2, shown in Fig. 1, and are further discussed here. The ACE gene was one of the first candidate genes thoroughly investigated for its possible relationship with the heart rate response to exercise [21]. Genetic association studies focusing on the effect of the ACE gene on heart rate increase during exercise reported many conflicting results [12, 22–25]. Some studies tested genes for their indirect effect on the sympathetic nervous system. One study observed that the NOS3 gene,

| Table 1 | Heritability estimates for the acute and long-term effect of exercise on heart rate response |
|---------|--------------------------------------------------------------------------------------------|
| Heritability type | Heritability | Type of exercise | Population | N | Author, year |
| Acute response: heart rate increase | | | | | |
| Family | 0.32 | Submaximal treadmill test | General population | 2053 | Ingelsson et al. (2007) [12] |
| SNP-based | 0.22 | Submaximal bicycle | General population | 58,818 | Verweij et al. (2018) [14] |
| SNP-based | 0.17 | Submaximal bicycle | General population | 66,665 | Ramirez J et al. (2018) [15] |
| Acute response: heart rate recovery | | | | | |
| Family | 0.34 | Submaximal treadmill test | General population | 2053 | Ingelsson et al. (2007) [12] |
| Twins and sibling | 0.60 and 0.65 | Maximal bicycle | General population | 491 | Nederend et al. (2016) [13] |
| SNP-based | 0.22 | Submaximal bicycle | General population | 58,818 | Verweij et al. (2018) [14] |
| SNP-based | 0.12 | Submaximal bicycle | General population | 66,665 | Ramirez et al. (2018) [15] |
| Trainings response: heart rate increase | | | | | |
| Family | 0.34 | Submaximal bicycle | General population | 481 | An et al. (2003) [17] |
| Family | 0.36 | Submaximal bicycle | Participants with high blood pressure | 529 | Rice et al. (2002) [16] |

Heritability estimates for the acute and long-term effect of exercise on heart rate response

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- 20 weeks during endurance training program at submaximal (50W) levels
Table 2  Summary of genes involved in acute heart rate increase

| Gene    | Variant | Chromosome/position | Minor/major allele/MAF | Type of study | Increase/decrease | P value         | Type of exercise test | Population | Author; year |
|---------|---------|---------------------|------------------------|---------------|-------------------|------------------|-----------------------|------------|--------------|
| ACE     | Del intron 16 | 17:63488529 Deletion/insertion | Candidate | – | > $5.00 \times 10^{-2}$ | Maximal and sub-maximal bicycle | General | Rankinen et al. (2000) [25] |
| ADRA1A  | rs544215 | 8:26712028 C/T/0.46 | Candidate | ↓ | $5.00 \times 10^{-3}$ | Standard Bruce | General | Ingelsson et al. (2007) [12] |
| ADRA1D  | rs3787441 | 20:4205059 G/A/0.27 | Candidate | ↓ | $7.00 \times 10^{-3}$ | Standard Bruce | General | Ingelsson et al. (2007) [12] |
| ADRB1   | rs1801253 | 10:114045297 C/G/0.28 | Candidate | ↓ | < $5.00 \times 10^{-2}$ | Maximal | Patients in cardiac rehab | Defoor et al. (2005) [29] |
| ADRB1   | rs1801252 | 10:114044277 G/A/0.21 | Candidate | ↓ | < $5.00 \times 10^{-2}$ | Maximal | Patients in cardiac rehab | Defoor et al. (2005) [29] |
| ADRB2   | rs1042713 | 5:148826877 A/G/- | Candidate | ↓ | < $5.00 \times 10^{-2}$ | Hand grip test | General | Eisenach et al. (2003) [30] |
| CAV2    | rs28495552 | 7:116113744 C/G/0.50 | GWAS | ↓ | $2.80 \times 10^{-11}$ | Sub-maximal bicycle | General | Ramirez et al. (2018) [15] |
| CCDC14I | rs10497529 | 2:179839888 A/G/0.04 | GWAS | ↑ | $2.50 \times 10^{-9}$ | Sub-maximal bicycle | General | Ramirez et al. (2018) [15] |
| GNAS1   | rs7121 | 20:58903752 C/T/0.37 | Candidate | ↑ | < $5.00 \times 10^{-2}$ | Ergometer | Referred for exercise test | Nieminen et al. (2006) [28] |
| HMG2    | rs1480470 | 12:66412130 A/G/0.37 | GWAS | ↑ | $3.40 \times 10^{-08}$ | Sub-maximal bicycle | General | Ramirez et al. (2018) [15] |
| MCTP2   | rs12906962 | 15:95312071 C/T/0.32 | GWAS | ↓ | $3.50 \times 10^{-13}$ | Sub-maximal bicycle | General | Ramirez et al. (2018) [15] |
| MCTP2   | rs12906962 | 15:95312071 C/T/0.33 | GWAS | ↓ | $2.70 \times 10^{-14}$ | Sub-maximal bicycle | General | Verweij et al. (2018) [14] |
| NOLA    | rs6847149 | 4:111157701 – | GWAS | – | $2.74 \times 10^{-06}$ | Standard Bruce | General | Vasan et al. (2007) [34] |
| NOS3    | rs1799983 | 7:150999023 T/G/0.26 | Candidate | ↓ | $4.00 \times 10^{-2}$ | Naughton protocol | Post-menopausal women | Hand et al. (2006) [26] |
| PAX2    | rs11190709 | 10:102552663 G/A/0.12 | GWAS | ↑ | $1.30 \times 10^{-11}$ | Sub-maximal bicycle | General | Ramirez et al. (2018) [15] |
| POP4    | rs12986417 | 19:30109533 A/G/0.35 | GWAS | ↓ | $1.00 \times 10^{-9}$ | Sub-maximal bicycle | General | Verweij et al. (2018) [14] |
| POP4    | rs7255293 | 19:30104198 G/A/0.42 | GWAS | ↓ | $3.20 \times 10^{-9}$ | Sub-maximal bicycle | General | Ramirez et al. (2018) [15] |
| PPIL1   | rs236352 | 6:36817113 A/G/0.34 | GWAS | ↑ | $6.40 \times 10^{-10}$ | Sub-maximal bicycle | General | Ramirez et al. (2018) [15] |
which produces nitric oxide, was associated with heart rate increase during exercise [26]. Although nitric oxide is mostly known for its vasodilatory effects, it is also thought to have a modulating effect on the parasympathetic and sympathetic nervous system [27]. GNAS1 was found to be associated with heart rate increase during exercise as well [28]. This gene encodes the G protein α-subunit that influences the sympathetic nervous system as it enables the coupling between adenyllyl cyclase and β1-adrenergic receptors. On the other hand, many studies brought forward genes based on their direct involvement in the sympathetic nervous system, and associations were found with the ADRB1 [29] and ADRB2 [30] genes, which both encode for β-adrenergic receptors. Interestingly, many previous findings could not be replicated in the Framingham Offspring study, which investigated multiple genes instead of focusing on a single gene. In this study, associations were found with the ADRA1A and ADRA1D genes [12]. These genes encode for α-adrenergic receptors that are mainly involved in smooth muscle cell contraction during sympathetic stimulation [12]. However, associations with the ADRB1 and ADRB2 genes could not be re-established [12].

| Table 2 (continued) | Gene | Variant | Chromosome/position | Minor/major allele/MAF | Type of study | Increase/decrease | P value | Type of exercise test | Population | Author; year |
|----------------------|------|---------|---------------------|------------------------|---------------|-------------------|---------|-------------------|------------|--------------|
| RGS6                 | rs17180489 | 14:72885471 | C/G/0.14           | GWAS                   | ↑              | 2.50×10^{-11}    |         | Sub-maximal bicycle | General   | Verweij et al. (2018) [14] |
| RNF220               | rs272564 | 1:45012273 | C/A/0.28           | GWAS                   | ↓              | 7.40×10^{-12}    |         | Sub-maximal bicycle | General   | Ramirez et al. (2018) [15] |
| RPL11                | rs58065122 | 8:10526186 | A/G/0.42           | GWAS                   | ↑              | 3.90×10^{-10}    |         | Sub-maximal bicycle | General   | Ramirez et al. (2018) [15] |
| RYR2^c,d             | rs2819770 | 1:234237045 | –                  | GWAS                   | –              | 3.53×10^{-6}     |         | Standard Bruce     | General   | Vasan et al. (2007) [34] |
| SCN10A               | rs7433723 | 3:38784957 | G/A/0.42           | GWAS                   | ↓              | 4.50×10^{-8}     |         | Sub-maximal bicycle | General   | Ramirez et al. (2018) [15] |
| SNCAIP               | rs4836027 | 5:121866990 | C/T/0.32           | GWAS                   | ↓              | 1.70×10^{-15}    |         | Sub-maximal bicycle | General   | Verweij et al. (2018) [14] |
| SNCAIP               | rs4836027 | 5:121866990 | C/T/0.31           | GWAS                   | ↓              | 9.90×10^{-21}    |         | Sub-maximal bicycle | General   | Ramirez et al. (2018) [15] |
| SOX5                 | rs4246224 | 12:24784139 | A/G/0.15           | GWAS                   | ↑              | 1.80×10^{-14}    |         | Sub-maximal bicycle | General   | Ramirez et al. (2018) [15] |
| SYT10                | rs1343676 | 12:33537387 | T/C/0.51           | GWAS                   | ↓              | 1.50×10^{-11}    |         | Sub-maximal bicycle | General   | Ramirez et al. (2018) [15] |
| TCF4                 | rs1125313 | 18:52859261 | C/A/0.50           | GWAS                   | ↑              | 3.90×10^{-9}     |         | Sub-maximal bicycle | General   | Ramirez et al. (2018) [15] |

Genes found to be associated with heart rate increase during exercise are shown in alphabetical order and are then ordered on the year published. Variation stands for either an SNP or deletion/insertion mutation. MAF stands for Minor Allele Frequency. Effects of a variant (in- or decrease) on heart rate increase during exercise are shown for the Minor Allele. Candidate stands for candidate gene study. GWAS stands for genome-wide association study. A hyphen is shown in case information which was not reported.

^Results from only one candidate gene study on ACE are shown; largest study was chosen; G allele in case of deletion; in case of insertion ATA CAGTCACTTTTTTTTTTTTTGGAGCATGCTCGCTGTCGCC

^Statistics from gene time of exercise interaction are showed

^Standard Bruce protocol is a maximal exercise treadmill test

^Statistics from the generalized estimating equations (GEE) tests are shown; Alleles were not mentioned in this article. None reached genome-wide significance; however, these were the most suggestive results

^Naughton protocol is a maximal exercise treadmill test
Although these studies were important for laying the foundation of our knowledge on the genetic determinants of heart rate increase during exercise, they failed to yield a comprehensive view by focusing on one or only a few genes. The Framingham Offspring study was the first to address these issues by conducting an early genetic linkage analysis on heart rate increase and recovery. However, not one genetic signal reached the appropriate significance level, which can possibly be attributed to the relatively low sample size of this study ($n = 2982$) [12]. In addition, linkage analyses have been shown to be less successful when applied to polygenic traits such as heart rate response to exercise [31], in part because of their limited power to detect the effect of common alleles with modest effects on disease [32].

More recently, genome-wide association studies (GWASs) were introduced. GWASs do have the potential to detect common alleles with modest effects on disease, since this method allows an unbiased and comprehensive search across the genome for single nucleotide polymorphisms (SNPs) [33]. The first GWAS on heart rate increase during exercise found $GAR1$ and $RYR2$ genes to be associated [34]. $GAR1$ is required for ribosome biogenesis and telomere maintenance. However, its specific function and how it possibly interacts with heart rate increase during exercise is unknown. $RYR2$ encodes a calcium channel that mediates calcium release from the sarcoplasmic reticulum into the cytoplasm and is therefore essential in triggering cardiac muscle contraction (Table 2, Fig. 1). $RYR2$ mutations in humans are associated with arrhythmogenic right-ventricular dysplasia and catecholaminergic polymorphic ventricular tachycardia. Interestingly, although caused by a different mutation in the $RYR2$ gene, both diseases are known to cause exercise-induced tachycardia [35–37]. However, these associations did not reach genome-wide significance, which might be due to the low sample size ($n = 1238$) [34].
Increasing the sample size for GWASs has been simplified by the development of inexpensive SNP arrays. Two GWASs were recently conducted on the acute heart rate response to exercise in the same cohort of the UK Biobank [14, 15]. The discussion of methodological differences between these studies has been published previously [38] and is beyond the scope of the current review. However, a summary of important differences is necessary to understand different genes found between the two studies. One difference is that the first study by Verweij et al. had a slightly lower sample size, since they used only echocardiography (ECG) measurements and did not include heart rate measurements derived by the UK Biobank itself. Another difference is that the study of Verweij et al. applied a more stringent threshold to claim a genome-wide significant level to be true (strategy to reduce the risk of type-1 errors) compared to the study published later by Ramirez et al. ($p < 8.3 \times 10^{-9}$ vs $p < 5.0 \times 10^{-8}$, respectively).

Of special interest are three genes that were found to be associated with heart rate increase during exercise in both studies, which are $\text{SNCAIP}$, $\text{MCTP2}$, and $\text{POP4}$ [14, 15]. The exact mechanism of $\text{SNCAIP}$ is not known so far; however, studies in mice have shown that $\text{SNCAIP}$ plays a role in neuronal degeneration (Table 2, Fig. 1) [39, 40]. $\text{POP4}$ is involved in the processing of precursor RNAs [41] and in the DNA damage response [42], thus preventing accumulation of deleterious mutations and DNA lesions and therefore potentially preventing genomic instabilities and carcinogenesis and prolonging neuronal life span. The $\text{MCTP2}$ gene is more specific to cardiac tissue. A mutation in the $\text{MCTP2}$ is known to cause left-ventricular outflow tract malformations in humans, which may alter the pressure within the ventricular outflow tract. Baroreceptors are densely located in this region and altered blood pressure could therefore lead to altered autonomic feedback on heart rate (Table 2, Fig. 1) [43]. Several other candidate genes found in these studies already provide a biological hypothesis to account for the associations with heart rate response to exercise. These genes can be broadly categorized into four categories, that is: (1) development of the nervous system, including the $\text{CCDC141}$ [44, 45], $\text{TCF4}$ [46, 47], $\text{PAX2}$ [48], $\text{SOX5}$ [49, 50], and $\text{CAV2}$ [51] genes; (2) prolongation of neuronal life span, including the $\text{SYT10}$ [52] gene; (3) cardiac development and disease, including $\text{RNF220}$ [53, 54] gene; and, finally, (4) genes involved in cardiac rhythm, including $\text{SCN10A}$ [55] and $\text{RGS6}$ [56, 57]. Of these, $\text{CCDC141}$, $\text{CAV2}$, $\text{SYT10}$, $\text{RNF220}$, and $\text{SCN10A}$ were more strongly associated with heart rate recovery after exercise (Tables 2, 3) and will be therefore discussed later. $\text{TCF4}$ is involved in the initiation of neuronal differentiation. Clinically, a mutation in $\text{TCF4}$ is known to cause Pitt-Hopkins syndrome, a severe congenital encephalopathy characterized by intellectual disability, developmental problems, seizures, breathing problems, and typical facial features [46, 47]. $\text{PAX2}$ encodes paired box gene 2 and is important in the early embryonic development as well. It is mostly known for its involvement in development of the kidney and urinary tract, since it is linked to papillorenal syndrome [58] and focal segmental glomerulosclerosis [59]. However, downstream target effectors of $\text{PAX2}$ have been hypothesized to be involved in neuronal development because of their supposed effect on the CHARGE syndrome [48]. $\text{SOX5}$ is involved in the regulation of chondrogenesis and the development of the nervous system [50]. In mice, it was found that loss of $\text{SOX5}$ resulted in decreased neuronal differentiation and secondary migrational abnormalities [49]. Mutations of the $\text{SOX5}$ gene in humans are known to cause the Lamb–Shaffer syndrome, which is characterized by speech delay, behavioural problems, and nonspecific dysmorphic features [50]. $\text{RGS6}$ is part of the regulation mechanism of the parasympathetic nervous system in the heart [56, 57]. It decreases muscarinic type 2 receptor (M2R) signalling in the sinoatrial node by rapidly terminating G$\beta$$\gamma$ signalling [56, 57]. In mice, it was shown that $\text{RGS6}$ knockdown removes the negative regulation of G$\beta$$\gamma$ leading to enhanced G protein-coupled inwardly rectifying potassium channel (GIRK)-induced sinoatrial and atrio-ventricular node hyperpolarization [56, 57]. It was therefore concluded that normal function of $\text{RGS6}$ is important for preventing parasympathetic override and severe bradycardia [56]. Its involvement in the parasympathetic nervous system was recently established in another GWAS in which it was found to be associated with heart rate variability [60], which is known to reflect parasympathetic activity [61]. Concerning heart rate increase during exercise, normal function of $\text{RGS6}$ probably facilitates parasympathetic withdrawal leading to the possibility to increase heart rate (Fig. 1).

Interestingly, none of the genes investigated in candidate gene studies were found to be associated with heart rate increase in any of the three GWASs. This is in line with the previous work in which early candidate gene studies were difficult to replicate [62, 63]. Two genes, $\text{HMGA2}$ and $\text{PPIL1}$, shown in Table 2 have not been discussed so far. $\text{PPIL1}$ is a gene that was recently found to be associated with heart rate variability as well [60]. However, to our knowledge, there is no current biological hypothesis to explain the association between $\text{PPIL1}$ or $\text{HMGA2}$ and heart rate increase during exercise.

**Heart rate recovery**

Heart rate recovery is characterized by increased parasympathetic tone followed by sympathetic withdrawal, which follows an inverse gradient pattern compared to heart rate increase [19]. It was elegantly shown in a dual-blockade study that especially parasympathetic reactivation is essential for interindividual differences in heart rate recovery.
| Gene       | Variant | Chromosome/position | Minor/major allele/MAF | Increase/decrease | $P$ value | Type of study | Type of exercise test | Time (seconds after exercise) | Population | Author; year |
|------------|---------|---------------------|------------------------|-------------------|-----------|---------------|-----------------------|-------------------------------|------------|--------------|
| ACE        | –       | –                   | Deletion/insertion     | –                 | $<1.00 \times 10^{-2}$ | Candidate | Running for 25 min with heart rate between 165 and 170 | 1800 | Athletes | Voroshin et al. (2008) [22] |
| ACHE       | rs3757868 | 7:100482720           | A/G/0.18               | ↓                 | $5.60 \times 10^{-24}$ | GWAS       | Submaximal bicycle | 40 | General | Verweij et al. (2018) [14] |
| ACHE       | rs3757868 | 7:100482720           | A/G/0.18               | ↓                 | $6.90 \times 10^{-11}$ | GWAS       | Submaximal bicycle | 50 | General | Ramirez et al. (2018) [15] |
| ADRA1B     | rs11953285 | 5:159324389              | C/A/0.13              | ↑                 | $1.00 \times 10^{-2}$ | Candidate | Standard Bruce | 180 | General | Ingelsson et al. (2007) [12] |
| ADRA2B     | del1301–303 | 2:96115249            | /AAGAGG AG/0.37       | ↓                 | $1.00 \times 10^{-2}$ | Candidate | Maximal bicycle | 60 | General | Kohli et al. (2015) [65] |
| ALG10B     | rs4533105 | 12:38214611            | C/T/0.43              | ↓                 | $1.90 \times 10^{-13}$ | GWAS       | Submaximal bicycle | 50 | General | Ramirez et al. (2018) [15] |
| BCAT1      | rs4963772 | 12:24758480            | A/G/0.15              | ↑                 | $1.20 \times 10^{-28}$ | GWAS       | Submaximal bicycle | 40 | General | Verweij et al. (2018) [14] |
| BCL11A     | rs1372876 | 2:60025963             | A/C/0.41              | ↓                 | $3.30 \times 10^{-9}$ | GWAS       | Submaximal bicycle | 50 | General | Ramirez et al. (2018) [15] |
| CAV2       | rs1997571 | 7:116198621            | G/A/0.48              | ↓                 | $1.70 \times 10^{-12}$ | GWAS       | Submaximal bicycle | 20 | General | Verweij et al. (2018) [14] |
| CAV2       | rs2109514 | 7:116159961            | A/G/0.50              | ↑                 | $7.10 \times 10^{-10}$ | GWAS       | Submaximal bicycle | 50 | General | Ramirez et al. (2018) [15] |
| CCDC141,TTN| rs17362588 | 2:179721046             | A/G/0.08              | ↓                 | $3.10 \times 10^{-9}$ | GWAS       | Submaximal bicycle | 10 | General | Verweij et al. (2018) [14] |
| CCDC141,TTN| rs35596070 | 2:179759692             | A/C/0.14              | ↓                 | $4.20 \times 10^{-13}$ | GWAS       | Submaximal bicycle | 10 | General | Verweij et al. (2018) [14] |
| CHRM2      | rs324640  | 7:136142625            | C/T/0.39              | ↓                 | $8.00 \times 10^{-3}$ | Candidate | Maximal bicycle | 60 | General | Hautala (2006) [66] |
| CHRM2      | rs8191992 | 7:136158563            | A/T/0.37              | ↓                 | $2.50 \times 10^{-3}$ | Candidate | Maximal bicycle | 60 | General | Hautala (2006) [66] |
| CHRM2d     | rs324640  | 7:136142625            | C/T/0.44              | ↓                 | $1.70 \times 10^{-3}$ | Candidate | Symptom-limited maximal bicycle | 60 | Post-MI | Hautala (2009) [67] |
| CHRM2d     | rs8191992 | 7:136158563            | A/T/0.43              | ↓                 | $1.60 \times 10^{-3}$ | Candidate | Symptom-limited maximal bicycle | 60 | Post-MI | Hautala (2009) [67] |
| CHRM2      | rs17168815 | 7:136624621            | T/G/0.16              | ↓                 | $1.10 \times 10^{-14}$ | GWAS       | Submaximal bicycle | 50 | General | Verweij et al. (2018) [14] |
| CHRM2      | rs6943656 | 7:13669436             | G/A/0.16              | ↓                 | $2.30 \times 10^{-10}$ | GWAS       | Submaximal bicycle | 50 | General | Ramirez et al. (2018) [15] |
| CLPB, INPPL1| rs7130652 | 11:71984398            | T/G/0.07              | ↑                 | $3.40 \times 10^{-11}$ | GWAS       | Submaximal bicycle | 10 | General | Verweij et al. (2018) [14] |
| Gene  | Variant   | Chromosome/position | Minor/major allele/MAF | Increase/decrease | P value     | Type of study | Type of exercise test | Time (seconds after exercise) | Population | Author; year |
|-------|-----------|---------------------|------------------------|-------------------|-------------|--------------|---------------------|-------------------------------|-------------|---------------|
| CNTN3 | rs34310778| 3:74783408          | C/T/0.43               | ↑                  | 1.00 × 10⁻⁹ | GWAS         | Submaximal bicycle   | 30                            | General     | Verweij et al. (2018) [14] |
| CNTN3 | rs6549649 | 3:74786491          | C/G/0.64               | ↑                  | 1.40 × 10⁻⁹ | GWAS         | Submaximal bicycle   | 50                            | General     | Ramirez et al. (2018) [15] |
| GNG11 | rs180238  | 7:93550447          | C/T/0.35               | ↓                  | 2.20 × 10⁻¹²| GWAS         | Submaximal bicycle   | 40                            | General     | Verweij et al. (2018) [14] |
| GRIK2 | rs2224202 | 6:102053814         | A/G/0.19               | ↑                  | 5.80 × 10⁻⁹ | GWAS         | Submaximal bicycle   | 20                            | General     | Verweij et al. (2018) [14] |
| KCNH8 | rs73043051| 3:18883863          | C/T/0.22               | ↑                  | 7.80 × 10⁻⁹ | GWAS         | Submaximal bicycle   | 50                            | General     | Verweij et al. (2018) [14] |
| MCTP2 | rs12906962| 15:95312071         | C/T/0.32               | ↓                  | 5.10 × 10⁻⁹ | GWAS         | Submaximal bicycle   | 50                            | General     | Ramirez et al. (2018) [15] |
| MED13L| rs61928421| 12:116227249        | T/C/0.07               | ↓                  | 4.30 × 10⁻¹⁵| GWAS         | Submaximal bicycle   | 40                            | General     | Verweij et al. (2018) [14] |
| MED13L| rs11067773| 12:11622895         | C/T/0.09               | ↓                  | 3.10 × 10⁻¹¹| GWAS         | Submaximal bicycle   | 50                            | General     | Ramirez et al. (2018) [15] |
| NDUFA11| rs12974440| 19:5894386          | A/G/0.08               | ↓                  | 2.40 × 10⁻¹⁰| GWAS         | Submaximal bicycle   | 10                            | General     | Verweij et al. (2018) [14] |
| NDUFA11| rs12974991| 19:5894584          | A/G/0.09               | ↓                  | 2.10 × 10⁻⁹ | GWAS         | Submaximal bicycle   | 50                            | General     | Ramirez et al. (2018) [15] |
| NEGR1 | rs61765646| 1:72723211          | A/T/0.19               | ↑                  | 1.10 × 10⁻¹³| GWAS         | Submaximal bicycle   | 10                            | General     | Verweij et al. (2018) [14] |
| PAX2  | rs4917911 | 10:102559421        | G/A/0.11               | ↑                  | 6.60 × 10⁻¹⁵| GWAS         | Submaximal bicycle   | 50                            | General     | Ramirez et al. (2018) [15] |
| PAX2  | rs7072737 | 10:102552663        | G/A/0.11               | ↑                  | 1.10 × 10⁻¹⁷| GWAS         | Submaximal bicycle   | 40                            | General     | Verweij et al. (2018) [14] |
| PRDM6 | rs151283  | 5:122446619         | A/C/0.28               | ↓                  | 1.60 × 10⁻¹⁰| GWAS         | Submaximal bicycle   | 50                            | General     | Verweij et al. (2018) [14] |
| PRKAG2e| rs1029947 | 7:150713400         | –                      | –                  | 9.20 × 10⁻⁷ | GWAS         | Standard Bruce⁵       | 180                           | General     | Vasan et al. (2007) [34]   |
| PRKAG2e| rs1029946 | 7:150713454         | –                      | –                  | 3.89 × 10⁻⁶ | GWAS         | Standard Bruce⁵       | 180                           | General     | Vasan et al. (2007) [34]   |
| RGS6  | rs150330648| 14:72844765          | T/G/0.01               | ↓                  | 4.30 × 10⁻⁸ | GWAS         | Submaximal bicycle   | 50                            | General     | Ramirez et al. (2018) [15] |
| RNF220| rs272564  | 1:45012273          | C/A/0.29               | ↓                  | 1.40 × 10⁻¹²| GWAS         | Submaximal bicycle   | 50                            | General     | Verweij et al. (2018) [14] |
| RNF220| rs272564  | 1:45012273          | C/A/0.28               | ↓                  | 8.80 × 10⁻¹⁰| GWAS         | Submaximal bicycle   | 50                            | General     | Ramirez et al. (2018) [15] |
| Gene   | Variant       | Chromosome/position | Minor/major allele/MAF | Increase/decrease | $P$ value     | Type of study | Type of exercise test | Time (seconds after exercise) | Population | Author; year                           |
|--------|---------------|---------------------|------------------------|-------------------|---------------|---------------|----------------------|-------------------------------|------------|---------------------------------------|
| SCN10A | rs6795970     | 3:38766675          | A/G/0.40               | ↓                  | $2.60 \times 10^{-8}$ | GWAS          | Submaximal bicycle   | 50                            | General   | Ramirez et al. (2018)                  |
| SERINC2| rs11589125    | 1:31894396          | T/C/0.06               | ↑                  | $6.60 \times 10^{-9}$ | GWAS          | Submaximal bicycle   | 50                            | General   | Verweij et al. (2018)                  |
| SKAP   | rs2158712     | 7:26582733          | T/A/0.48               | ↓                  | $2.80 \times 10^{-13}$ | GWAS          | Submaximal bicycle   | 10                            | General   | Verweij et al. (2018)                  |
| SNCAIP | rs1993875     | 5:121869310         | C/G/0.30               | ↓                  | $9.50 \times 10^{-9}$ | GWAS          | Submaximal bicycle   | 50                            | General   | Ramirez et al. (2018)                  |
| SOX5   | rs112630705   | 12:24773919         | A/G/0.15               | ↑                  | $3.20 \times 10^{-11}$ | GWAS          | Submaximal bicycle   | 50                            | General   | Ramirez et al. (2018)                  |
| SYT10  | rs6488162     | 12:33593127         | T/C/0.42               | ↓                  | $2.60 \times 10^{-66}$ | GWAS          | Submaximal bicycle   | 10                            | General   | Verweij et al. (2018)                  |
| SYT10  | rs2218650     | 12:33734783         | A/G/0.15               | ↓                  | $1.10 \times 10^{-26}$ | GWAS          | Submaximal bicycle   | 50                            | General   | Ramirez et al. (2018)                  |

Genes found to be associated with heart rate recovery after exercise are shown in alphabetical order and are then ordered on the year published. Mutation stands for either a SNP or deletion/insertion mutation. MAF stands for Minor Allele Frequency. Effect of variant (increase or decrease) on heart rate increase during exercise is shown for the minor allele. Candidate stands for candidate gene study. GWAS stands for genome-wide association study. MI stands for myocardial infarction. A hyphen is shown in case information which was not reported.

*a* Insignificant after correction for multiple testing  
*b* Standard Bruce protocol was used  
*c* Heart rate recovery constant was used as measurement. This fits heart rate recovery to a first-order exponential decay curve  
*d* Compared lowest and highest HRR quartile. Linear model showed no significant association  
*e* Statistics from the family-based association tests are shown; Alleles were not mentioned in this article. None reached genome-wide significance. Results shown are the strongest associations.
However, the exact mechanisms underlying these differences remain to be determined. Twin, family, and GWA studies estimated the genetic component to interindividual differences of heart rate recovery after one minute to range between 0.12 and 0.60 (Table 1). Therefore, genetic studies may yield novel insights into heart rate recovery. All genetic determinants investigated for their potential causal role in interindividual differences in heart rate recovery are summarized in Table 3 and are discussed below. An illustration of possible causal genes and how they are supposed to influence acute heart rate recovery after exercise is shown in Fig. 2.

Initially, the same candidate genes were proposed for heart rate recovery as for heart rate increase. For example, the ACE gene was found to be related to heart rate recovery in one candidate gene study as well [22]. Another study found ADRA1B and ADRA2B to be associated with heart rate recovery (Table 3) [12]. The association between ADRA2B gene and heart rate recovery was also found in another candidate gene study [65]. Other studies focused primarily on the parasympathetic nervous system represented by the CHRM2 gene. The minor alleles of the rs324640 and rs8191992 SNPs found in CHRM2 region were found to be associated with a lower heart rate recovery in the general population [66] and in patients with a history of myocardial infarction [67]. In addition, these minor alleles increased chances of death to coronary artery disease in the latter group [67].

The problem of biased selection of candidate genes has been solved by conducting GWASs as previously stated. The first GWAS on the acute heart rate response to exercise found heart rate recovery measured 3 min post-exercise to be associated with PRKAG2, though this association did not reach genome-wide significance. PRKAG2 is involved in the regulation of ATP restoration after periods of ATP depletion.

Fig. 2 Graphical representation of genes (shown in italic) involved in acute heart rate recovery after exercise grouped by working mechanism (shown in bold). The left and left upper part of the figure shows the nervous system. The middle upper part zooms in on a parasympathetic neuron of the vagus nerve (twice) and its synapse. Note that although we zoom in on the brain stem (which is the main location of parasympathetic nuclei that innervate the vagus nerve), we actually show a peripheral parasympathetic neuron of the vagus nerve. The heart is displayed on the right; the upper right of the figure shows the aorta with next to it a pacemaker cell in the cardiac sinus node. In the middle of the figure, below, we zoom in on cardiac tissue and receptors. Cholinergic receptors and enzymes are shown in light blue and glutamate receptors in yellow. Sodium and potassium channels are shown in red and pink, respectively.
and therefore might influence the return of heart rate to its initial state (Table 3, Fig. 2).

As previously mentioned, sample size was drastically increased in the two recent studies in the UK Biobank [14, 15]. Some differences between both studies have been discussed earlier (i.e., sample size and genome-wide significant threshold). Concerning heart rate recovery, it is worth mentioning that the phenotype definition was not equal between both studies. The study of Ramirez et al. [15] determined heart rate recovery traditionally as the difference between maximum heart rate and heart rate approximately 1 min after cessation of exercise. The study of Verweij et al. defined heart rate recovery at five time points, which included the differences between maximum heart rate and heart rate after 50, 40, 30, 20, and 10 s after exercise. This includes heart rate recovery at earlier time points (i.e., 10 s), which was recently established to be a superior predictor of outcome of all-cause mortality and death by coronary artery disease [6, 7].

Interestingly, both studies found the previously investigated candidate gene CHRM2 to be associated with heart rate recovery [14, 15]. CHRM2 encodes M2R, the main muscarinic cholinergic receptor in the heart. This receptor is known for both its negative chronotropic and inotropic effects after binding with acetylcholine released by post-ganglionic parasympathetic nerves [69]. An increase of acetylcholine is known for both its negative chronotropic and inotropic effects after binding with acetylcholine released by post-ganglionic parasympathetic nerves (Table 3, Fig. 2) [68].

The role of the parasympathetic nervous system in interindividual differences in heart rate recovery is additionally highlighted by the ACHE gene that was found in both studies. ACHE encodes for acetylcholinesterase, an enzyme which breaks down acetylcholine in the synaptic cleft of postganglionic parasympathetic nerves [69]. An increase of acetylcholinesterase would therefore cause an attenuated heart rate recovery by decreasing parasympathetic reactivation. Other genes that were found in both studies were SYT10, CNTN3, PAX2, CAV2, MED13L, RNF220, and NDUFAL1 (Table 3, Fig. 2). SYT10 encodes a Ca$^{2+}$ sensor synaptotagmin 10 that triggers IGF-1 exocytosis, which, in turn, protects neurons from degeneration. SYT10 might play an important role in the regulation of heart rate, as it was found to be associated with resting heart rate [3, 23], heart rate increase [15], and heart rate variability [60] as well. CNTN3 belongs to a group of glycosylphosphatidyl-anchored cell adhesion molecules that are mostly found in neurons [70, 71]. Because of its similarity with TAG-1, it is thought to have an important function in neuronal outgrowth and wiring of the nervous system [70–72]. In the study of Ramirez et al. it was found that the allele of one SNP decreased heart rate recovery and increased CNTN3 expression levels in the nucleus accumbens [15]. Since heart rate recovery is mainly influenced by the parasympathetic nervous system [64], it was hypothesized that CNTN3 may also be relevant to cardiac parasympathetic modulation [15]. However, it is more likely to be associated with cardiac sympathetic modulation, since morphology of the nucleus accumbens has been shown to be correlated with cardiac sympathetic index [73]. PAX2 is known to be the first gene to be expressed in the mid- and hindbrain during embryonal developments in mice [74] and can be found in the hindbrain in the early stages of embryo development in humans as well [48]. The hindbrain includes the nucleus tractus solitarius, nucleus ambiguous, and dorsal nucleus of the vagus, which are known to mainly influence cardiac parasympathetic innervation of the heart through vagus nerve stimulation [75]. Less is known about CAV2, which was found to be associated with heart rate response to exercise as well. However, one study pointed out that CAV2 is necessary for differentiation of dorsal root ganglion cells during the early differentiating programs [51]. The function of MED13L is unclear as well, but knockdown in zebrafish caused abnormal neural-crest cell migration [76]. This is supported by clinical characteristics in humans with MED13L mutations, which can be characterized by intellectual disabilities, developmental delay, and craniofacial anomalies [77]. RNF220 functions as an E3 ubiquitin ligase, which determines protein target specificity during posttranslational ubiquitination [53]. A possible link with heart rate recovery originates from the involvement of RNF220 in the canonical WNT signalling cascade. In a knockdown study, RNF220 was shown to stabilize β-catenin by interacting with ubiquitin-specific peptidase Usp7 [54]. This stabilizing function is important, because the WNT/β-catenin signalling pathway is involved in embryonic cardiac development [78], the development of cardiac disease [79–81], and in cardiac repair [80]. NDUFA11 is an accessory subunit of the mitochondrial membrane respiratory chain NADH dehydrogenase complex I. In humans, a splice-site mutation in this gene is known to cause mitochondrial complex I deficiency. This can cause a wide range of disorders, including encephalomyocardioopathy [82]. Recently, it was shown that downregulation of NDUFA11 by small interfering RNA reduced ATP production and increased mitochondria reactive oxygen species production in cardiac mitochondria of mice [83]. NDUFA11 was found to be associated with heart rate variability as well, suggesting that it is an important factor in causing differences between individuals’ heart rate response [60].

Other candidate genes found in one of the GWASs provide a biological hypothesis for their possible causal role in interindividual differences in heart rate recovery as well. These genes include CDC14I, BCL11A, KCNH8, ALG10B, GNG11, GRIK2, and NEGR1. CDC14I is a gene that plays a central role in neuronal development [44, 45]. In fact, in utero knockdown of CDC14I in mice resulted in impaired radial migration in [44]. The same applies to BCL11A, which encodes a C2H2-type zinc-finger protein that is involved in neuronal development.
Studies in mice have shown that slowed migration of neurons upon knockdown resulted in microcephaly with decreased brain volume [84], particularly affecting the limbic system [85]. Within the human brain, it is most highly expressed in the caudate nucleus followed by hippocampus [86]. In humans, different de novo heterozygous mutations have been found to cause developmental disorder with persistence of fetal haemoglobin [85]. KCNH8 encodes a voltage-gated potassium channel. It is mainly expressed in the central nervous system and is involved in the regulation of neuronal excitation (Table 3, Fig. 2) [87–89]. ALG10B is involved in potassium regulation, as well, since it is a potassium channel regulator that couples to KCNH2. However, it is more involved in cardiac tissue than neuronal tissue and is known for its influence on heart rhythm. Upon binding with KCNH2, it reduces sensitivity to classic proarrhythmic drug blockade [90]. GNG11 encodes the γ11 subunit of the heterotrimeric G protein complex Goβγ [91]. GNG11 is just as RGS6 thought to be involved in GIRK activation and was found to be associated with heart rate variability [60] as well. In this study, it was hypothesized that variations in this gene lower the availability of the γ11 subunit, thereby reducing Goβγ component-induced GIRK activation [60]. This would lead to decreased heart rate variability through attenuated response to changes in cardiac vagal activity [60]. If true, the same would apply for heart rate recovery; decreased response to cardiac vagal reactivation after exercise would translate to blunted heart rate recovery. In addition, another mutation in the RGS6 gene in humans was shown to decrease susceptibility to the long QT syndrome [92]. GRIK2 encodes a glutamate receptor that is mostly expressed in the human cerebral and cerebellar cortices [93]. Here, it is involved in neuronal excitation and plays an important role in a variety of normal neurophysiologic processes. Neuronal Growth Regulator 1 (NEGR1) is essential for neuronal morphology and, just as CNTN3, has been shown to regulate neurite outgrowth (Table 3, Fig. 2) [94]. Perhaps because of this essential function, NEGR1 has been associated with many polygenic traits, including body mass index, years of education, and physical activity.

Heart rate increase and recovery share a high genetic correlation and it is therefore likely that there is overlap in genes that were found for both aspects of the heart rate response to exercise [14]. SNCAIP, SOX5, RGS6, and MCTP2 genes were already discussed for heart rate variability [95] and heritability analysis estimated a genetic component ranging between 0.34 and 0.36 (Table 1) [16, 17]. Therefore, several studies were conducted to gain insights in the causes of these interindividual differences. The first study in the HERITAGE family cohort found a heritability of 0.34 for exercise heart rate changes to regular training, with the strongest linkage on chromosome 2q33.3-q34 [17]. Next, this region was fine-mapped and it was found that the CREB1 gene locus was strongly associated with submaximal exercise heart rate training response [96]. Nonetheless, it only explained 5.45% of the 34% heritability [96].

To gain further insights in the genes causing the remaining fraction of its heritability, a GWAS was performed in the HERITAGE family cohort. In this study, nine SNPs were identified and accounted for the total of 34% heritability of exercise-induced changes to heart rate increase [97]. The most significantly associated SNP was linked to the YWHAQ gene (Table 4). YWHAQ is mostly expressed in the brain, heart, and pancreas [98], and its main function is apoptosis and cell proliferation. It was shown that the cardiac-specific mutated YWHAQ gene leads to increased pathological ventricular remodelling with increased cardiomyocyte apoptosis after experimental myocardial infarction [99]. It can be hypothesized that mutations in the YWHAQ gene lead to similar pathological cardiac remodelling after exercise training, causing diminished exercise-induced changes to heart rate increase. However, a neurological causal pathway cannot be ruled out, since the same mechanism could apply to neuronal remodelling needed to attenuate heart rate increase after regular exercise training [11].

The CREB1 gene (Table 4) was significantly associated with submaximal heart rate response to exercise training as well [97]. In this study, it was hypothesized that CREB1 altered the exercise-induced changes in heart rate increase due to its effect on either cardiac [100] or neuronal memory [101]. Cardiac memory is a phenomenon in which an altered T wave on electrocardiogram can be seen when sinus rhythm restarts after a period of abnormal rhythm, for example, after ventricular pacing or arrhythmia [100]. The other hypothesis involving neuronal memory fits in our current understanding that neuron biology is of great importance in the heart rate response
Neuronal memory or long-term potentiation is a form of synaptic plasticity in which there is a long-lasting increase of synaptic strength in case the synapse is highly active. It could be hypothesized that regular exercise causes an increase of synaptic strength of parasympathetic neurons, thus altering the heart rate increase during exercise. However, CREB1 encodes a transcription factor that regulates many mechanisms in the body and its association with memory does not imply causality. A recent editorial rightfully addressed the fact that the same allele in another study was found to increase the rise of temperature and, therefore, might decrease subjective liking of exercise training, potentially diminishing motivation.

### Table 4 Summary of genes involved in the long-term heart response to exercise

| Gene   | Variant | Chromosome/position | P value | Type of study | Training schedule | Type of exercise test | Population | Author; year |
|--------|---------|---------------------|---------|---------------|-------------------|----------------------|------------|--------------|
| Heart rate increase |          |                     |         |               |                   |                      |            |              |
| CREB1a | rs2253206 | 2:208100223          | 1.6 × 10⁻⁵ | GWAS          | 20 weeks, 3 times a day, 30–50 min at submaximal HR | Submaximal bicycle | Healthy, but sedentary | Rankinen et al. (2012) [97] |
| GCH1a  | rs2057368 | 14:54373759          | 5.6 × 10⁻⁵ | GWAS          | 20 weeks, 3 times a day, 30–50 min at submaximal HR | Submaximal bicycle | Healthy, but sedentary | Rankinen et al. (2012) [97] |
| GPRIN3a | rs1560488 | 4:90444858           | 3.3 × 10⁻⁵ | GWAS          | 20 weeks, 3 times a day, 30–50 min at submaximal HR | Submaximal bicycle | Healthy, but sedentary | Rankinen et al. (2012) [97] |
| RBPMSa | rs2979481 | 8:30382328           | 3.8 × 10⁻⁶ | GWAS          | 20 weeks, 3 times a day, 30–50 min at submaximal HR | Submaximal bicycle | Healthy, but sedentary | Rankinen et al. (2012) [97] |
| MYLIPa | rs909562 | 6:16238312           | 3.2 × 10⁻⁵ | GWAS          | 20 weeks, 3 times a day, 30–50 min at submaximal HR | Submaximal bicycle | Healthy, but sedentary | Rankinen et al. (2012) [97] |
| OR6N2a | rs857838 | 1:157017174          | 7.6 × 10⁻⁵ | GWAS          | 20 weeks, 3 times a day, 30–50 min at submaximal HR | Submaximal bicycle | Healthy, but sedentary | Rankinen et al. (2012) [97] |
| PIWIL1a | rs4759659 | 12:129403241         | 5.7 × 10⁻⁵ | GWAS          | 20 weeks, 3 times a day, 30–50 min at submaximal HR | Submaximal bicycle | Healthy, but sedentary | Rankinen et al. (2012) [97] |
| TFECa  | rs10248479 | 7:115395591          | 3.4 × 10⁻⁵ | GWAS          | 20 weeks, 3 times a day, 30–50 min at submaximal HR | Submaximal bicycle | Healthy, but sedentary | Rankinen et al. (2012) [97] |
| YWHAQa | rs6432018 | 2:9639347            | 8.1 × 10⁻⁷ | GWAS          | 20 weeks, 3 times a day, 30–50 min at submaximal HR | Submaximal bicycle | Healthy, but sedentary | Rankinen et al. (2012) [97] |
| Heart rate recovery |          |                     |         |               |                   |                      |            |              |
| CHRM2b | rs324640 | 7:136146251          | 0.008   | Candidate     | 2 weeks, 5 times a day, 40 min at submaximal HR | Maximal bicycle | Healthy, but sedentary | Hautala (2006) [66] |
| CHRM2b | rs8191992 | 7:136158563          | 0.005   | Candidate     | 2 weeks, 5 times a day, 40 min at submaximal HR | Maximal bicycle | Healthy, but sedentary | Hautala (2006) [66] |

Genes found to be associated with changes in training-induced changes to heart rate increase and recovery are shown in alphabetical order. Variant stands for either a SNP or deletion/insertion mutation. Candidate stands for candidate gene study. GWAS stands for genome-wide association study.

a Allele frequencies and betas are not mentioned in this study and direction (increase or decrease of response to training) can, therefore, not be determined.

b Minor alleles of rs324640 and rs8191992 (respectively, A and C) decreased heart rate recovery.

to exercise. Urinary memory or long-term potentiation is a form of synaptic plasticity in which there is a long-lasting increase of synaptic strength in case the synapse is highly active. It could be hypothesized that regular exercise causes an increase of synaptic strength of parasympathetic neurons, thus altering the heart rate increase during exercise. However, CREB1 encodes a transcription factor that regulates many mechanisms in the body and its association with memory does not imply causality. A recent editorial rightfully addressed the fact that the same allele in another study was found to increase the rise of temperature and, therefore, might decrease subjective liking of exercise training, potentially diminishing motivation.
Heart rate recovery

On the other hand, heart rate recovery increases when the cardiac autonomic balance shifts towards vagal dominance after regular endurance training [10]. Little research has been performed on the genetics of training-induced changes to heart rate recovery, although a heritable component has been suggested [66]. To our knowledge, only one study has been conducted on this subject. In this candidate gene study, it was found that the CHRM2 gene (Table 4) is linked to long-term modification of heart rate recovery to exercise as well [66]. Participants who had a the minor alleles of the rs324640 and rs8191992 SNPs were not only found to have a lower acute heart rate recovery, but also showed less increase in heart rate recovery after regular endurance training. As previously mentioned, CHRM2 encodes the muscarinic acetylcholine receptor M2R and, upon activation, causes a negative chronotropic and inotropic response. It therefore seems that genetic variation in CHRM2 not only causes interindividual differences in acute heart rate recovery [68], but also in long-term modifications. A full overview of the genes discussed for the long-term heart rate response to exercise can be found in Table 4.

Association of heart rate response to exercise-related genes with other traits

We assessed the association of described genes with other traits in publicly available GWASs using the GWAS catalogue (Online Resource 1). In short, the candidate causal genes that were associated with both heart rate increase and recovery were also associated with resting heart rate (CCDC141, RGS6, RNF220, SCN10A, and SYT10), heart rate variability (CCDC141, RGS6, RNF220, and SYT10), blood pressure (CCDC141 and PAX2), atrial fibrillation (CAV2 and SCN10A), coronary artery disease (CAV2 and SCN10A), and ECG traits including the PR interval (CAV2 and SCN10A), QRS duration, and the Brugada syndrome (both SN10A).

Some genes that were only associated with heart rate increase during exercise were found to be associated with resting heart rate and heart rate variability (PPIL1), blood pressure (ADRB1, ACE, NOS3, and HMG2A), atrial fibrillation (MCTP2 and NOS3), exercise treadmill test and lung function (both RYR2). Similarly, some of the heart rate recovery genes were also associated with resting heart rate (ACHE and GNG11), heart rate variability (GNG11 and NDUFA11), blood pressure (PRDM6, PRKAG2, and CHRM2), QRS duration, (PRDM6), atrial fibrillation and coronary artery disease (BCL11A, PRDM6), and obesity and vigorous physical activity levels (both NEGR1).

Future directions

Improvement of prevention and treatment of disease in the human health sector is the ultimate application of novel knowledge found by genetic studies and future research should be performed to achieve this goal (Fig. 3) [104]. Functional follow-up of findings obtained by GWAS will be necessary to gain insights in how likely causal genes affect the heart rate response to exercise [104]. Most genes that were prioritized so far have a plausible biological mechanism in which they influence the heart rate response to exercise. However, the exact effect of all genes on exercise-induced heart rate changes could be validated in an experimental setting (Fig. 3). One possible method is to perform functional experiments in cardiomyocytes obtained from embryonic stem cells [105]. In cardiomyocytes, human diseases and risk factors with their underlying genetic contribution can be created in vitro [105]. Since cardiomyocyte cell cultures can beat spontaneously [105], simulating the effect of this genetic contribution allows for investigation of the acute heart rate response to pacing from resting to exercise heart rate levels in small cell cultures. In addition, by simulating the effect of this genetic contribution, drugs can be screened against an individual’s full genetic backgrounds to discover information on cardiotoxicity for each individual. This could potentially give insights in the development of personalized medicine strategies for heart rate modification [106], which is an essential strategy in the treatment of coronary artery disease [107] and heart failure (Fig. 3) [108, 109]. Genes known to affect cardiac de- and repolarisation (RYR2, ALG10B, and SCN10A) or GIRK channels in the cardiac sinus node (RGS6 and GNG11) could be of interest to study in this setting. Recent development in the generation of spinal human cord neural cells could provide the same opportunity for investigating neuronal cell longevity including genes such as SCA1P, POP4, and SYT10 [110]. Complex neurological mechanisms at the interplay of the sympathetic and parasympathetic nervous system (i.e., KCNH8 and GRIK2) or neuronal development (i.e., SOX5, PAX2, and BCL11A) are more difficult to investigate using this method. This can be solved by investigating these genes using in vivo models of animals that share a high percentage of their genomic pattern with humans, including mice [111, 112], fruit flies [113], and zebrafish [114] (Fig. 3). For example, knockdown of RGS6 [56, 57], MED13L [76], and BCL11A [85] has already provided insights in biological consequences of mutations in these genes.

Improvement of prevention of disease is another goal of genetic research. While accurate risk prediction might be relatively straightforward for mono- or oligogenic disease, this is more difficult for polygenic diseases such as coronary artery disease and heart failure. However, the
knowledge on genetic variants obtained by GWAS can be used to construct genetic risk cores by summing the number of risk alleles weighted by the corresponding beta coefficients. Recently, it was shown that the polygenic risk score of coronary artery disease had the ability to identify 8.0% of the population at greater than threefold risk for coronary artery disease [115]. These individuals can subsequently be selected for encouragement of behavioural lifestyle changes as relative effects of poor lifestyle were shown to be comparable between genetic risk groups [116]. Similar to the traditional risk score models in which several traditional risk phenotypes are used to predict risk events, this could ultimately be performed for genetic risk score models as well. As previously stated, there is a large body of observational studies that links heart rate response to exercise to all-cause mortality and cardiovascular disease in healthy individuals and those with a history of cardiac disease [5–9]. In this light, it would be interesting to see whether adding the polygenetic risk scores for the acute heart rate response to exercise into a genetic risk score model that includes the polygenetic risk score for the cardiovascular disease itself could improve detection of individuals at high risk of disease. However, it should be noted that both recent GWAS on the acute heart rate response to exercise did not find support for a genetic association with cardiovascular mortality [14, 15]. The lack of an association in both studies might originate from the fact that a small replication cohort consisting of a relatively young and healthy population was used. The study of Verweij et al. [14] did find a significant association between heart rate response to exercise and parental age as proxy for all-cause mortality. However, first, it is required to investigate whether there is a genetic association with cardiovascular disease and all-cause mortality, preferably in a larger independent cohort [117].

The evidence on long-term modification of the heart rate response to exercise is limited so far [97]. If the genetics of the acute heart rate response to exercise can be used to predict cardiovascular mortality, the combination with information on the genetics of the long-term modification of the heart rate response to exercise could one day inform the choice of prevention strategy. For example, a high genetic risk score for a diminished acute response to exercise combined with a genetic...
risk rate response could be an indicator of early primary or secondary prevention strategies (Fig. 3). On the other hand, a high genetic risk score for a diminished acute response to exercise combined with a genetic risk score that indicates little training-induced changes could be an indication of early intervention through medication (Fig. 3).

**Conclusion**

In the current review, we found a total of 10 genes associated with the acute heart rate response to exercise in candidate gene studies. Only one gene (CHRM2), related to heart rate recovery, was replicated in recent GWASs. Additional 17 candidate causal genes were identified for heart rate increase and 26 for heart rate recovery in these GWASs. Nine of these genes were associated with both acute heart rate increase and recovery during exercise. These genes can be broadly categorized into four categories: (1) development of the nervous system (CCDC141, PAX2, SOX5, and CAV2); (2) prolongation of neuronal life span (SYT10); (3) cardiac development (RNF220 and MCTP2), and (4) cardiac rhythm (SCN10A and RGS6). Of the total of 43 genes, nine showed overlap with resting heart rate and heart rate variability, six with atrial fibrillation and coronary artery disease, two with ECG traits, and nine with blood pressure. The current findings support the idea that the autonomic nervous system is a major player in the regulation of the acute heart rate response to exercise. Heart rate recovery is especially influenced by parasympathetic nervous system genes (ACHE and CHRM2), in line with the previous research [64]. Regarding the long-term response to exercise, heart rate increase during exercise was found to be mainly associated with genes involved in either cardiac or neuronal remodeling. Little evidence has been found for the long-term response of heart rate recovery to exercise, except for parasympathetic involvement. Future work will be required to translate these findings to preventive and therapeutic applications.

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**Compliance with ethical standards**

**Conflict of interest** The authors declare that they have no conflict of interest. Author N. Verweij is an employee of Genomics plc.

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