

1 The Cardiovascular Disease Knowledge Portal: A community resource for cardiovascular
2 disease research
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68 Non-standard Abbreviations and Acronyms

69 CVD: cardiovascular disease

70 CVDKP: Cardiovascular Disease Knowledge Portal

71 HuGeAMP: Human Genetics Amplifier

72 HF: heart failure

73 CAD: coronary artery disease

74 AF: atrial fibrillation

75

76 Better understanding the mechanisms underlying cardiovascular disease (CVD) is critical for
77 improving prediction, diagnosis, and treatment of these conditions affecting millions of people
78 worldwide¹. Large-scale genetic association studies have elucidated the genetic architecture of
79 CVD biomarkers and outcomes, yet translating these data into biological knowledge is
80 challenging. To assist with this challenge, we developed the Cardiovascular Disease Knowledge
81 Portal (CVDKP; broadcvdi.org) in collaboration with CVD researchers. The CVDKP is built on
82 the Human Genetics Amplifier (HuGeAMP)² and is a component of the Common Metabolic
83 Diseases Knowledge Portal (cmdkp.org) along with portals focused on cerebrovascular disease³,
84 diabetes², and sleep disorders.

85

86 The CVDKP contains three dataset types. First, it includes 238 genetic association datasets for
87 173 phenotypes related to CVD, including results from the HERMES (heart failure, HF),
88 CARDIoGRAMplusC4D (coronary artery disease, CAD), MiGen (myocardial infarction),
89 AFGen (atrial fibrillation, AF), ICBP (blood pressure), GLGC (plasma lipids), and GIANT
90 (anthropometric traits) consortia. In selecting genetic datasets, we prioritize: (a) CVD-relevant
91 phenotypes not yet included in the CVDKP; (b) newer, larger datasets for major phenotypes; and
92 (c) studies of non-European populations (to increase the diversity of ancestries represented in the

93 CVDKP). Second, the CVDKP contains >5,500 tissue-specific epigenomic annotations (*e.g.*,
94 predicted cis-regulatory element locations), including nearly 350 for heart tissues and blood
95 vessels. Finally, it includes lists of predicted “effector genes” for CAD, AF, HF, blood pressure,
96 and plasma lipids, curated from publications that that prioritize causal genes at genetically
97 associated loci. Although the CVDKP is funded by the Accelerating Medicines Partnership® in
98 Common Metabolic Diseases (which includes pharmaceutical industry partners), decisions about
99 datasets and content are made by academic CVD researchers and are not biased toward drug
100 discovery. Datasets are biased, however, towards common rather than rare conditions due to the
101 availability and focus of large-scale genetic association studies.

102
103 We apply a series of five bioinformatic methods (described fully in Costanzo *et al.*²) to these
104 datasets to support high-level queries. We use a “bottom-line” meta-analysis method to compute
105 a consensus p-value for a variant across all datasets, accounting for sample overlap; dataset-level
106 associations are also viewable for each variant. This analysis eliminates associations in one
107 dataset that are unsupported by other datasets (for example, 202 out of the 597 “dataset-level”
108 associations across CAD, AF, and HF). It also uncovers novel associations that only become
109 significant after meta-analysis (30 such associations across CAD, AF, and HF), including
110 rs12209223 for AF, which lies within an intron of Filamin-A-Interacting Protein 1 (*FILIP1*), a
111 gene expressed in skeletal muscle, cardiac tissue, and arteries. To highlight epigenomic
112 annotations, pathways, or other traits related to a disease of interest, we apply MAGMA pathway
113 analysis across all gene sets in the Molecular Signatures Database, stratified LD-Score regression
114 across all epigenomic datasets, and cross-trait LD-Score regression across all pairs of traits. As
115 illustrated for AF (**Figure**, panel A), these analyses together capture its genetic relationship to

116 disease causes (high BMI) and consequences (heart failure), underlying disease mechanisms
117 (*e.g.*, atrioventricular blockage, cardiac muscle cell membrane depolarization), and tissues
118 through which genetic associations exert their effects (*e.g.*, heart, cardiovascular system, muscle
119 structure).

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121 To access these results, the CVDKP contains four main pages, centered on a variant, gene,
122 genomic region, or phenotype. Each page contains visualizations of associations and
123 bioinformatic method results for the selected object and can be filtered to show ancestry-specific
124 results when available. The portal also contains a menu of tools, each of which implements a
125 multi-step filter or analysis. Some visualizations are intended for biologists unfamiliar with
126 human genetics⁴: for example, the gene page includes a high-level summaries of a gene's
127 “genetic support” (**Figure**, panels B and C), that omits technical details of association analyses.
128 Other visualizations are intended for researchers well-versed in human genetics: for example, the
129 tools menu includes a workflow that mimics widely used “variant to function” analyses (**Figure**,
130 panel D) and programmatic access to all results. The CVDKP also draws information about gene
131 product function and pathways from external resources: UniProt (<https://www.uniprot.org/>),
132 MyGene (<https://mygene.info/>), and MSigDB (<https://www.gsea-msigdb.org/gsea/msigdb/>).

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134 The goal of the CVDKP is to provide a disease-specific resource that includes the genomic data
135 types, datasets, and methods considered authoritative by the CVD research community. Its focus
136 on genomic data complements other CVD-focused resources, such as HeartBioPortal⁵ and
137 BioDataCatalyst. We make the CVDKP accessible to all researchers – regardless of geographical
138 location or career stage – by offering it as an open-access site with educational resources for all

139 learning styles, including written and video documentation, participatory webinars and focus
140 groups, and a responsive helpdesk. We welcome new collaborations with the research
141 community as we continue to develop the CVDKP.

142

143 All data and materials have been made publicly available at the CVDKP and can be accessed at
144 <https://broadcvdi.org/>. The meta-analysis results from published GWAS datasets are not
145 considered to be human subjects research and therefore do not require IRB approval.

146

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199 **Figure legend**

200 Scientific questions that can be addressed using the CVDKP. **A:** The Phenotype page, illustrated
201 for the “Atrial fibrillation” (AF) phenotype, illuminates the genetic architecture of a disease or
202 trait by displaying (top to bottom) a Manhattan plot of bottom-line genetic associations
203 accompanied by a table (not shown) of the top associations; a plot and table of common variant
204 gene-level associations and a table of pathways enriched for the top genes associated with AF,
205 both determined using the MAGMA algorithm (de Leeuw CA *et al.*, PLoS Comput Biol. 2015;
206 11:e1004219); and a table of phenotypes whose genetic architecture is correlated with that of
207 AF, as determined by cross-trait LD-Score regression (Finucane HK *et al.*, Nat Genet. 2015;
208 47:1228). **B:** Top, the Region page near the *MYO9B* gene shows significant associations for
209 coronary artery disease (CAD), total cholesterol, and diastolic blood pressure. Bottom, scores
210 from the Human Genetic Evidence calculator on the Gene page, which takes into account variant
211 impact, common variant associations, and rare variant associations to generate summaries of the
212 evidence for involvement of a gene in a disease or trait, predict that *MYO9B* may have roles in

213 CAD and traits that impact CAD risk. This gene has been experimentally shown to have a role in
214 regulating vascular cell motility, supporting a role in CAD risk (Aragam K *et al.*, Nat Genet.
215 2022; 54: 1803). **C:** Lists of predicted effector genes, curated from the literature, are available
216 for multiple diseases and traits; the Gene page links to any effector lists on which that gene is
217 represented. The *MYO9B* gene (see panel A) is included in 8 effector lists, for CAD, blood
218 pressure, and lipid traits. Shown, part of the interactive table representing CAD effector gene
219 predictions from the CARDIoGRAMplusC4D consortium (Aragam K *et al.*, Nat Genet. 2022;
220 54:1803), which considered 8 types of evidence to predict the causal gene at each CAD-
221 associated genetic locus. The table includes summary rows that are expandable (bottom) to show
222 detailed evidence. **D.** The Variant Sifter displays genetic associations, credible sets, tissue-
223 specific epigenomic annotations, and variant-gene links determined from chromatin
224 conformation assays, as well as an interactive table (not shown) listing coordinates and
225 parameters and linking to the source datasets. In this example, the LDL-associated variant
226 rs2618566 near the *RRBP1* gene is seen to have high posterior probability in LDL credible sets,
227 to be located in a predicted enhancer region in HepG2 cells, and to be linked to the *RRBP1*
228 promoter in both adipose and liver tissue, recapitulating findings by Ramdas *et al.* (Am J Hum
229 Genet. 2022;109:1366).