Understanding cardiovascular disease: a journey through the genome (and what we found there)

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Cardiovascular disease (CVD) is a major cause of mortality and hospitalization worldwide. Several risk factors have been identified that are strongly associated with the development of CVD. However, these explain only a fraction of cases, and the focus of research into the causes underlying the unexplained risk has shifted first to genetics and more recently to genomics. A genetic contribution to CVD has long been recognized; however, with the exception of certain conditions that show Mendelian inheritance, it has proved more challenging than anticipated to identify the precise genomic components responsible for the development of CVD. Genome-wide association studies (GWAS) have provided information about specific genetic variations associated with disease, but these are only now beginning to reveal the underlying molecular mechanisms. To fully understand the biological implications of these associations, we need to relate them to the exquisite, multilayered regulation of protein expression, which includes chromatin remodeling, regulatory elements, microRNAs and alternative splicing. Understanding how the information contained in the DNA relates to the operation of these regulatory layers will allow us not only to better predict the development of CVD but also to develop more effective therapies.

“We shall not cease from exploration
And the end of all our exploring
Will be to arrive where we started
And know the place for the first time.”
Little Gidding by T. S. Eliot (1942)

Introduction
Cardiovascular disease (CVD) is the leading cause of morbidity and mortality worldwide, responsible for an estimated 17.5 million deaths in 2005, representing 30% of all deaths (Mudd and Kass, 2008). In Europe alone, CVD causes over 4.3 million deaths each year, and is the leading cause of death (48%) and disease burden (23%). Despite substantial advances in medical management, the prognosis of CVD remains poor, and the identification of mechanisms and potential therapeutic approaches are still a priority of considerable importance.

CVD has a very clear environmental component; however, the risk factors defined in epidemiological studies (e.g. hypertension, high cholesterol, smoking) explain only a fraction of events (Thanassoulis and Vasan, 2010). Much hope has been set on the potential of genetics and genomics to reveal the molecular mechanisms responsible for the development of CVD and to explain cases that are not obviously correlated with known risk factors. The strong familial component of CVD has long been recognized and is evident from the large-scale population studies that helped to define the classical set of CVD risk factors. However, studies of CVD heritability are confounded by the fact that several other risk factors, such as blood pressure, lipid levels and diabetes, are themselves under genetic control (North et al., 2003). Nonetheless, several studies have noted that family history is an independent risk factor (Shea et al., 1984; Myers et al., 1990), so that CVD heritability does not merely reflect the genetic component of classical risk factors. Unfortunately, evidence of a high degree of heritability of a given trait does not mean that identification of the underlying genes will be straightforward. The analysis of common multifactorial diseases such as CVD is hindered by the interdependence of genetic and environmental factors and the difficulties that are inherent in separating the influence of individual factors.

In order to fully grasp the contribution of genetics to CVD, we must go beyond classical Mendelian genetics, and explore the multiple interacting layers that regulate the genome. This will include the analysis of not only the protein-coding sequences of the genome, but the vast non-coding regions as well. An unbiased view of the whole genome is now possible, thanks to the development of large-scale approaches led by next generation sequencing (NGS) technologies. (For a Glossary of terms used, see Box 1.) In this review, we discuss how studying the many layers regulating the genome – including chromatin structure, cis-regulatory elements, microRNAs and alternative splicing – is beginning to provide us with a fresh view of the development of and therapeutic strategies for CVD.

The past: understanding the heritability of CVD
To date, most knowledge about genes causing CVD comes from the study of congenital heart defects and rare forms of familial
inherited CVD. Congenital heart defects can be non-syndromic or syndromic (such as Holt-Oram and DiGeorge syndromes), and usually affect genes involved in the early phases of heart development; these genes can encode transcription factors (such as Gata4, Nkx2.5 or Tbx factors), signaling pathway components (such as those of the Notch or the Ras-MAPK pathways) or structural proteins (such as the α-myosin heavy chain encoded by the MYH6 gene) (Srivastava, 2006). Cardiac developmental defects are the most common human birth defects (Hoffman and Kaplan, 2002) and their study remains of great importance. Furthermore, a still-uncharted territory is that of developmental defects that do not cause an overt fetal phenotype but that subtly impair proper cardiac function or stress responses and are an underlying morphological substrate for later-onset CVD.

Familial conditions are usually caused by autosomal dominant mutations that show a variable degree of expressivity. Identification of the culprit genes underlying these inherited CVDs has allowed genetic testing in families with affected members, therefore allowing counseling and prevention in asymptomatic carriers (Tester and Ackerman, 2011). Inherited heart diseases are broadly classified into cardiomyopathies and channelopathies. In cardiomyopathies, such as hypertrophic cardiomyopathy or arrhythmogenic right ventricular cardiomyopathy, there is a structural alteration in the heart (Watkins et al., 2011). By contrast, the hearts of individuals with channelopathies are structurally normal, but these conditions, such as long QT or Brugada syndromes, mainly cause arrhythmias and can result in sudden cardiac death (Bastaiaen and Behr, 2011).

Although information gained from the study of these and other conditions has been extraordinarily useful in defining the genetic basis of cardiovascular morphology and physiology, they only account for a very small fraction of heritable CVDs. The existence of these conditions has also prompted numerous studies in which candidate genes, identified because of their developmental function in animal models or in tissue culture assays, have been screened for disease association in affected families or small case-control studies. However, many associations identified in this way have not been replicated, showing the inherent limitations of this approach (Ioannidis et al., 2001; Morgan et al., 2007).

#### The present: GWAS, eQTLs and exome sequencing

Genome-wide association studies (GWAS), which have the potential to fill the gap in our understanding of the genetic basis of CVD and other common diseases (Hirschhorn and Daly, 2005). Because they examine the entire genome in an unbiased fashion, GWAS approaches can identify any genomic region involved in a given disease and should therefore be free of any kind of ascertainment bias, which is the main problem with candidate gene approaches. Overall, there is no doubt that GWAS have provided interesting new biological insights into many conditions. Surprising findings include the discovery of SNPs in genes that were originally not thought to have a role in disease and the identification of loci shared by diseases, including CVDs, that were previously thought to be unrelated (Pandey, 2010; O’Donnell and Nabel, 2011).

However, GWAS have, in the view of many, failed to deliver as expected. The main criticisms are that the identified alleles explain only a small fraction of the heritability of common diseases and traits (Manolio et al., 2009) and have a low predictive value compared with classical risk factors (Thannouslis and Vasan, 2010). Possible reasons for this apparent failure are that initial expectations were too high or that the wrong questions were posed. In addition, because GWAS are designed to identify common variants, the existence of rare variants with a large effect has not been addressed. Consequently, initial hopes that the identification of a small number of SNPs would explain the inheritance of CVD and other diseases and serve as predictors, similarly to classical Mendelian inheritance, have not been realized, because this is not the aim of GWAS.

GWAS methodology has improved and possible explanations for missing heritability are currently being explored (Zuk et al., 2012); moreover, ongoing investigation of GWAS results continues to increase the number of variants associated with common traits. A major strategy developed to overcome the limitations of GWAS is a brute-force attack by means of meta-analysis, in which information from multiple GWAS analyses is pooled to increase the number of cases and controls analyzed (Cantor et al., 2010; Pandey, 2010). Another strategy to refine and complement GWAS is targeted re-sequencing of disease-associated loci to identify rare or causal variants. This approach has successfully identified rare
variants associated with hypertriglyceridemia (Johansen et al., 2010), although exactly how the rare and common variants interact and associate with the disease is still unclear.

Despite the limitations of GWAS analyses, their unbiased searching of the whole genome for associations has identified previously unknown genetic components of CVD. Some of the discovered loci were completely unexpected, opening new avenues of research into the pathways and processes underlying disease. For example, variants discovered at the 9p21 region were the first common genetic variants identified as genetic risk factors for coronary artery disease and other forms of CVDs independent of classical risk factors (Burton et al., 2007; Helgadottir et al., 2007; McPherson et al., 2007; Samani et al., 2007). The SNP variants in the 9p21 gene desert region constitute an early and elegant example of how disease association can be related to molecular phenomena. Different mechanisms for this association have been proposed, such as disruption of regulatory elements involved in the interferon-γ signaling response (Harismendy et al., 2011) or of a non-coding RNA (Pasmant et al., 2011). Another unexpected association is that of SORT1, encoding the multi-ligand sorting protein sortilin 1, with plasma low-density-lipoprotein cholesterol levels and myocardial infarction; this association has uncovered a newly identified pathway involved in lipid metabolism and CVD, thus offering new possibilities for therapeutic intervention (Dube et al., 2011).

It is also notable that many disease-associated variants identified by GWAS are located in loci that have previously been linked to Mendelian diseases (Lupski et al., 2011). For example, mutations in KCNE1, encoding a potassium channel subunit, are responsible for congenital long QT syndrome (Splawski et al., 2000) and Jervell and Lange-Nielsen syndrome, and GWAS-identified variants near this gene are associated with QT interval duration (Newton-Cheh et al., 2009). A similar finding has been reported for SCN5A, which is associated with Brugada syndrome (Ruan et al., 2009). Therefore, classical Mendelian genetics and some GWAS converge on a similar range of genomic loci, the only difference being the nature and effect of the variants (Lupski et al., 2011).

Gene expression and eQTLs

Gene expression in heart disease has been studied for a long time. Some genes, such as NPPA [which encodes atrial natriuretic factor (ANF)], NPPB [brain natriuretic peptide (BNP)], ACTA1 (α-skeletal actin) and MIH7 (β-myosin heavy chain), are known to be induced in stressed cardiomyocytes in infarcted, hypertrophic and dilated hearts. Indeed, some of these genes have been used for diagnosis and prognosis in heart disease.

Gene expression in the heart is strongly heritable (Petretto et al., 2006), which suggests that it is genetically controlled. By combining microarray studies with genetic linkage analysis we can identify control points in the genome that regulate gene expression. These are considered expression quantitative trait loci (eQTL) and can help reveal genetic associations for certain traits for which GWAS has not provided a clear answer. One such trait is left ventricular mass, a risk factor for heart failure and a predictor of all-cause mortality. By using microarray and linkage analysis, an association was found, both in rats and humans, between increased left ventricular mass and elevated expression of the extracellular matrix protein osteoglycin (Petretto et al., 2008). After hypertrophic stimulation, osteoglycin knockout mice show reduced ventricular mass compared with wild-type mice, establishing a causal relationship between osteoglycin expression and increased left ventricular mass. The same group recently established an association between ventricular mass and endonuclease G (Endog), a mitochondrial nuclease that controls mitochondrial mass (McDermott-Roe et al., 2011). Loss-of-function mutations in Endog result in increased left ventricular mass and a decline in cardiac function in rats. Deletion of the Endog gene in mice induces mitochondrial depletion and dysfunction, increased oxidative stress, and cardiac steatosis and hypertrophy. eQTL analysis has also unveiled an inflammatory network, driven by interferon regulatory factor 7 (Irf7), that is mainly expressed by monocytes and macrophages and is associated with susceptibility to type 2 diabetes (Heinig et al., 2010). Expression of the genes in this network correlates with genetic variation in Epstein-Barr virus-induced gene 2 (Ebi2), suggesting that the network is controlled by this gene. These studies reveal how eQTL analysis can improve our knowledge of the biological basis of CVD.

Exome sequencing

With the advent of NGS technologies, exome sequencing – the targeted sequencing of the portion of the human genome that is protein coding – has become a widely used tool for discovering rare alleles underlying Mendelian phenotypes and complex traits. Although recent examples abound of the successful use of exome sequencing to identify genes underlying monogenic disorders (Bamshad et al., 2011), this approach has yet to make a major contribution to our understanding of complex diseases, and is currently the focus of a major debate. Large-scale sequencing efforts have shown that putative loss-of-function variants in protein-coding genes are abundant in apparently healthy individuals (MacArthur and Tyler-Smith, 2010; Conrad et al., 2011), which calls for caution when establishing causal relations between mutations identified by exome sequencing and disease etiology. In the cardiovascular field, exome sequencing has identified several previously unknown candidate causal genes for Mendelian disorders, including SHROOM3 for heterotaxy (Tariq et al., 2011) and ANGPTL3 in cases of familial combined hyperlipidemia (Musunuru et al., 2010). In other cases in which exome sequencing failed to identify the causal variant, complementary approaches such as copy-number variation analysis did so; for example, variation in BAG3 causes familial dilated cardiomyopathy (Norton et al., 2011). Exome sequencing and what the future holds for this technology have recently been reviewed (Bamshad et al., 2011). However, it is likely that this approach will soon be superseded by whole genome sequencing, which has the advantage of being free of the ascertainment bias inherent to exome sequencing, which explores just 1% of the genome.

Layers of regulation in CVD

Genomic analysis has so far fallen short of expectations in terms of establishing the underlying genetic cause of CVD. Although associations between certain loci and disease have been found, their biological impact is not immediately obvious because we remain largely ignorant of the role of genomic elements located outside coding regions.

Whole individual genomes will be fully sequenced in the near future, and comparison of these sequences might identify new
associations with CVD. However, it is arguable whether these data will yield biological information about the molecular mechanisms that underlie CVD. Even if this information is encoded in the DNA sequence and can be found, it is by no means certain that we would be able to understand it. The production of a particular protein in a cell from the information encoded in DNA is a process that is exquisitely controlled; intertwined layers of regulation provide both qualitative and quantitative control of gene expression. To understand the real contribution of the different elements in the genome to CVD, it is crucial to carry out a multilayered analysis of cells and tissues. Such an approach must include the analysis of chromatin modification and epigenetic regulation, activation of regulatory elements, gene transcription, alternative splicing and non-coding RNAs, among other levels of control (Fig. 1). We therefore need to look beyond the DNA sequence and integrate different layers of biological information so that we can return to the starting point, look again at the DNA sequence and, hopefully, understand for the first time what is written in it (as prefigured in Eliot’s poem).

Chromatin modifications
DNA is wound around histones to form chromatin, which is in turn tightly packed to save space. For gene transcription to occur, chromatin needs to be locally unwound to make the locus accessible to the transcription machinery. Different modifications along the chromatin thread determine whether it will be open or remain closed. These are mainly acetylation and methylation modifications of histones and DNA, which can be inherited from cell to cell through mitosis and constitute an epigenetic level of gene regulation. Epigenetic regulation is particularly sensitive to environmental changes and is thought to be a major mechanism by which external stimuli induce an inheritable response (Ordovas and Smith, 2010). In addition, differences in epigenetic modifications might explain changes in disease susceptibility in the absence of DNA sequence variation. Epigenetic changes are not written in the DNA sequence but can be detected at a global level through ChIP-seq, which combines immunoprecipitation of modified histones linked to DNA with NGS. A recent report describes distinct epigenetic modifications identified in end-stage heart failure patients (Movassagh et al., 2011).

Epigenetic modifications are carried out by families of histone acetyl transferases (HATs), histone deacetylases (HDACs), histone methyl transferases (HMTs) and histone demethylases (HDMs). These protein families play a major role in heart development and disease (Ohtani and Dimmeler, 2011). Histone-modifying enzymes do not bind to DNA but are recruited by transcription factors, coactivators and repressors that provide the sequence specificity that these enzymes lack. Histone acetylation occurs on lysine...
residues and promotes DNA unwinding to facilitate transcription, whereas deacetylation has the opposite effect. The role of HDACs in heart disease is diverse and has been thoroughly studied in animal models (Haberland et al., 2009). HDAC9 and HDAC5 control cardiac growth, and knockout mice for Hdac9 are more sensitive to hypertrophic stimuli (Zhang et al., 2002; Chang et al., 2004). Furthermore, variants in the vicinity of HDAC9 are implicated in causing increased risk of large vessel ischemic stroke (Bellenguez et al., 2012). HDAC2 contributes to the development of cardiac hypertrophy and re-expression of the fetal gene program associated with cardiac stress (Trivedi et al., 2007). HDAC3 is necessary for proper heart metabolism. Cardiac-specific deletion of the gene encoding this protein results in deregulation of PPARγ, lipid accumulation and massive cardiac hypertrophy. HATs also play roles in heart disease; for example, alterations in the levels of p300 can lead either to developmental cardiac defects or to hypertrophy and heart failure (Yao et al., 1998; Miyamoto et al., 2006; Wei et al., 2008). Although these and other studies highlight roles for HATs and HDACs in heart development and disease, it should be noted that these enzymes might have targets other than histones and that their effects might not necessarily be directly linked to chromatin modification.

Histones can be methylated on different lysine residues and can undergo different degrees of methylation, resulting in distinct effects on transcription regulation (Ohtani and Dimmeler, 2011). Loss of K4 (Lys4) trimethylation on histone 3 (H3) deregulates expression of ion channels and cytoskeleton genes and results in altered contraction (Stein et al., 2011). Like HDACs, HDMs play a major role in heart development and disease. In the embryo, the HDM Jarid2 regulates cardiomyocyte proliferation and is necessary for proper heart development (Lee et al., 2000; Toyoda et al., 2003). Another HDM, UTX, acts on trimethylated K27 to de-repress transcription of cardiac genes and switch on cardiac differentiation (Lee et al., 2012). In the adult heart, expression of JMJD2A is increased in individuals with hypertrophic cardiomyopathy and plays a key role in the development of cardiac hypertrophy, as revealed through gain- and loss-of-function mouse models (Zhang et al., 2011). The role of HMTs in the heart is less well understood. A recent report shows that the HMT DotL1 controls dystrophin transcription and that DotL1 knockout mice develop dilated cardiomyopathy (Nguyen et al., 2011). For a full review of the role of these proteins in cardiovascular development, see Chang and Bruneau (Chang and Bruneau, 2012).

Chromatin is also remodeled by the Brg1/Brm-associated factor (BAF) complexes in an ATP-dependent manner. The BAF complex is composed of several proteins, some of which have been reported to play a major role in the heart. For example, Brg1 expression is induced in individuals with hypertrophic cardiomyopathy and is necessary for cardiac hypertrophy development in mice (Hang et al., 2010). Another BAF-complex component, Baf60c, plays a major role in cardiac differentiation and heart development (Lickert et al., 2004). Together, these reports highlight how chromatin-remodeling regulators control gene expression and susceptibility to CVD.

Cis-regulatory elements
The identification and characterization of the functional sequence elements that determine when, where and how much a gene is expressed is a major goal in biology (Birney et al., 2007; Myers et al., 2011). Acquiring this knowledge will explain a large fraction of the genetic component of CVD. However, at present this is a daunting task because we still have not learned to recognize regulatory elements based on sequence alone (as can be done for coding regions), and their identification thus far relies on time-consuming small-throughput functional assays. Furthermore, there is still no clear estimate of how many functional elements are present in the genome, although it is expected that a large fraction of non-coding sequences (the vast majority of the genome) will contain such elements. The early observation (Dermitzakis et al., 2002) that sequence conservation of non-coding regions is comparable to that of coding regions further supports a functional role for non-coding regions, because they have been conserved throughout evolution and subjected to positive selection.

A major frustration with GWAS has been that the great majority of risk-associated SNPs are in the non-coding portion of the genome, and usually at a great distance from the nearest protein-coding gene or in different linkage disequilibrium blocks. This disappointment arises from the common view that coding variants will have a more profound effect than those in functional elements present in the non-coding regions (Cooper and Shendure, 2011). Nevertheless, there is little direct evidence supporting this, and we are only starting to glimpse the complexity of gene regulation through non-coding functional elements. Because these vastly outnumber coding exons, the abundance of non-coding risk variants identified by GWAS should come as no surprise.

The advent of new technologies such as ChiP-seq that allow genome-wide screening to identify regions bound by specific proteins (transcription factors) or that show certain epigenetic features (DNA methylation, nucleosome-free regions or histone modifications) has opened new avenues in the search for functional cis-regulatory elements. Recent work has shown how combinations of these epigenetic marks distinguish active from inactive chromatin regions, or proximal regulatory elements (promoters) from distal elements (Heintzman et al., 2007; Rada-Iglesias et al., 2011). Active regulatory elements have also been shown to possess an ‘open’ chromatin configuration characterized by depletion of nucleosomes, which can be identified by DNase I hypersensitivity or FAIRE (formaldehyde-assisted identification of regulatory elements) assays (Song et al., 2011). With these tools in hand, it is now possible to generate a catalog of candidate regulatory elements in distinct genomic regions.

Investigators have begun to use such approaches to determine the regulatory basis of heart development and function, with the expectation that such knowledge will yield useful insights into the genetic component of CVD. Analysis of the binding of the co-activator protein p300 to chromatin has identified thousands of putative heart enhancers in mice (Blow et al., 2010) and humans (May et al., 2011). More importantly, functional testing in mouse transgenic assays has validated the predicted enhancer function of a large proportion of these predicted elements. Other studies using a combination of bioinformatic, conservation and functional validation have also predicted an extensive list of genomic locations with putative regulatory activity in the heart (Narlikar et al., 2010).

In other cases, the genome-wide binding profiles of cardiac-specific transcription factors (including Gata4, Nkx2.5, Tbx5, Srf and Mef2a) have been defined using in vitro mouse tissue culture systems (He et al., 2011; Schlesinger et al., 2011). These studies showed that
these factors often bind to the same genomic regions, identifying cis-regulatory modules, in a similar manner to that described in other systems such as embryonic stem cells (Chambers and Tomlinson, 2009). The integration of these data with maps of histone modification, DNA methylation, co-activator and nucleosome exclusion in cardiovascular tissues will provide a comprehensive view of how the genome is dynamically regulated in health and disease.

Nevertheless, the generation of complete catalogs of cis-regulatory elements presents some problems. First, functional validation of predictions based on genome-wide scans will be only as good as the assays used. At present, robust assays for enhancer function are available (both for tissue culture and in vivo transgenesis). However, novel assays to test for silencer, insulator or other functions need to be developed and applied in a systematic fashion to classify predicted elements. Second, once these elements have been found, it is necessary to identify which genes they act on. Although common sense would predict that the nearest gene will be the target, this is not necessarily the case. To address this issue, high-throughput chromatin interaction maps can be built to show the genomic architecture in a tissue-specific manner (de Wit and de Laat, 2012). And last but not least, a major problem facing studies of tissue-specific patterns of gene regulation is the availability of samples, both from patients and controls. This is particularly the case for the study of CVD, because biopsies are rare and in many circumstances the etiology of the diseased tissue is complex and multifactorial. Attempts have been made to correlate changes in gene expression and regulation by using another easily accessible tissue as a proxy (such as blood). However, careful analysis shows that gene expression patterns are not correlated even between closely related samples (Powell et al., 2011).

**Alternative splicing**

After the completion of the human genome project, researchers were somewhat puzzled, and fairly disappointed, by the low number of protein-coding genes in the genome. The 21,000 genes in the genome must suffice to make the ~1 million proteins estimated to be present in the human proteome. Protein diversity originates not so much in DNA, but in RNA. As genes are being transcribed, introns are removed from the immature mRNA and exons are linked together in a process known as mRNA splicing. Rather than being a rigid process, splicing allows alternative combinations of exons that result in the production of different proteins from a single gene. In fact, alternative splicing (AS) is considered the main factor underlyng protein diversity.

Virtually all genes (94%) with more than one exon undergo AS. AS events include exon inclusion or exclusion, mutually exclusive exons, usage of alternative 5’ or 3’ splice sites and intron retention. These changes result in alterations in the final mRNA product that can lead to shifts in the open reading frame, generation of premature stop codons (nonsense-mediated decay) or changes in protein domains. Between 50% and 80% of splicing events are regulated in a tissue-specific manner (Wang and Burge, 2008). A recent study using exon microarrays showed reduced splicing efficiency and altered AS of sarcomeric genes in heart-failure patients (Kong et al., 2010). However, our knowledge about the involvement of AS in heart pathophysiology is mainly limited to individual genes whose different isoforms play diverse roles during heart failure. For example, AS variants of the troponin I gene reduce contraction efficiency, and AS-generated titin isoforms alter cardiac stiffness in individuals with dilated cardiomyopathy (Makarenko et al., 2004; Feng and Jin, 2010). AS of the sodium channel SCN5A, which mediates cardioprotection by ischemic preconditioning, generates two non-functional variants during heart failure (Shang et al., 2007). Similarly, a splicing variant of cell-cycle-regulated kinase (CCRK), which promotes cardiomyocyte growth and survival, is downregulated in heart failure (Qiu et al., 2008). We have also observed that the CnAβ1 AS variant of the phosphatase calcineurin improves cardiac function after infarction instead of inducing maladaptive hypertrophy like other calcineurin isoforms (Felkin et al., 2011).

AS is regulated by cis-regulatory genomic sequences that are present in the alternatively spliced exon and its flanking introns, and by trans-regulatory factors that recognize these sequences. The role of some of these trans-regulatory factors has been studied in animal models. Knockout mice for SF2/ASF, a splicing enhancer of the serine and arginine-rich (SR) protein family, show contraction defects that lead to diluted cardiomyopathy and death within the first 8 weeks of life due to postnatal developmental defects (Xu et al., 2005). Similarly, knockout of the SR protein SC35 leads to dilated cardiomyopathy 3-5 weeks after birth, although SC35 itself is not necessary for cardiac development (Ding et al., 2004). Elevated levels of the CELF trans-regulatory factor CUGBP1 are associated with myotonic dystrophy, and overexpression of a dominant-negative CELF protein results in cardiac hypertrophy, fibrosis and dilated cardiomyopathy (Ladd et al., 2005; Wang et al., 2007). Interestingly, Rbm20 knockout mice were recently reported to develop dilated cardiomyopathy due to defects in titin splicing (Guo et al., 2012). These defects are also detected in humans with Rbm20 mutations. Although these reports highlight the general importance of AS trans-regulatory factors in embryonic and postnatal development, their role in the response of the heart to pathological stimuli and their therapeutic potential has barely been explored.

**Non-coding RNAs**

Much of what was once considered – somewhat presumptuously – ‘junk’ DNA has turned out to be actively transcribed to produce RNAs that do not encode protein information [non-coding RNAs (ncRNAs)] but instead act as regulators of other RNAs. Depending on their size, origin or function, ncRNAs are classified as microRNAs, long non-coding RNAs (lncRNAs) and so on. Although these elements are transcribed from intergenic or intronic regions of the genome, they might hold the key to understanding how genetic variation can impact protein expression.

microRNAs are endogenous small non-coding RNAs of ~22 nucleotides that constitute the predominant form of double-stranded RNA (dsRNA) in mammalian cells. Unlike other interference RNAs, most microRNAs do not induce mRNA cleavage, but instead silence protein expression through post-transcriptional mechanisms, either preventing mRNA translation or regulating its stability. Rather than an on-off switch, microRNAs can be viewed as biological rheostats that fine-tune the expression of a specific protein (Bartel and Chen, 2004; Baek et al., 2008). MicroRNA-encoding genes are usually found in intergenic, intronic or polycistrionic regions and lack a canonical TATA box and introns (Rana, 2007). MicroRNAs are transcribed in the nucleus mainly by type II RNA polymerases and are processed first by...
Drosha and then by Dicer into a mature microRNA (Rana, 2007). Depending on its degree of complementarity with the microRNA, the target mRNA will either be prevented from undergoing translation or will be degraded (Rana, 2007). Target mRNAs often carry more than one microRNA-recognition sequence in their 3’UTR and can sometimes be regulated by two different microRNAs. In addition, each microRNA can target up to hundreds of different mRNAs, the efficiency depending on the complementarity between microRNA and mRNA (Baek et al., 2008). MicroRNAs play a major role in CVD, which has been discussed extensively elsewhere (Latronico and Condorelli, 2009).

lncRNAs play diverse — and somewhat unexpected — regulatory roles (Wilusz et al., 2009). Long intergenic non-coding RNAs (lincRNAs) are transcribed from enhancer elements and act as transcriptional enhancers for neighboring genes (De Santa et al., 2010; Ørom et al., 2010), a role also ascribed to some short ncRNAs called eRNAs (Kim et al., 2010). The mechanism of action of these lncRNAs is not entirely clear, although they are speculated to facilitate the assembly of histone-modifying complexes (Ong and Corces, 2011). An additional category of lncRNAs was recently described, called competing endogenous RNAs (ceRNAs) (Franco-Zorrilla et al., 2007; Tay et al., 2011). ceRNAs mimic mRNA target sequences for microRNAs, resulting in competition between ceRNA and mRNA for the microRNA. In this way, ceRNAs act as microRNA ‘sponges,’ regulating the availability of microRNAs and indirectly controlling the amount of protein that will be translated from the mRNA target. Although nothing is known about ceRNAs in the heart, the ceRNA line-MD1 was recently identified as a regulator of muscle differentiation by sequestering miR-133 and miR-135, which results in increased expression of their targets (MAML1 and MEF2C) and activation of muscle-specific transcription (Cesana et al., 2011). More than 100 lncRNAs were shown to be differentially expressed in hypertrophic mouse hearts in a recent RNA-seq analysis (Lee et al., 2011). However, their general function and specific roles in heart disease remain unknown.

**Intertwined regulation**

It is important to note that all of the layers of biological regulation discussed above also control each other (Fig. 1). For instance, both chromatin modification and microRNAs modulate AS (Kalsotra et al., 2010; Luco et al., 2010). This regulation is bidirectional: splicing activation can promote the recruitment of HM Ts, and changes in AS regulatory sequences result in the repositioning of epigenetic marks (de Almeida et al., 2011; Kim et al., 2011). Like protein-coding genes, ncRNAs are epigenetically regulated and some of them can in turn modulate chromatin modification (Szulwach et al., 2010). As explained above, ncRNAs can also act as transcriptional enhancers. Together, these reports clearly indicate that the expression of a protein in a given cell and during a given time frame is very delicately regulated. If we are to infer information from DNA sequences, we need to integrate all available data on the genome together with correlative studies with other -omics analyses (such as transcriptomics, metabolomics, exposomics), which will provide a broader conceptual framework within which to understand the hereditability of common diseases. The characterization of the majority of these genomic features will require functional assays in tissue culture or in vivo that will have to be carried out in animal models of CVD (e.g. mouse, rat, dog, pig). Another approach that holds great promise is the generation of patient-specific induced pluripotent stem (iP) cells, which can then be differentiated towards a cardiovascular phenotype. Such ‘disease-in-a-dish’ models have begun to be used for the study of Mendelian CVD (Carvajal-Vergara et al., 2010; Itzhaki et al., 2011; Yazawa et al., 2011) and population-scale GWAS analysis in iP-cell-derived cardiac cells is not far away.

**The future**

With sequencing costs dropping rapidly, we will be soon shift from analysis of exomes to genomes, enabling us to interrogate many more variants that might be important for controlling gene transcription, splicing or even translation. Technical challenges remain, but work to surmount these difficulties is progressing rapidly. For instance, the limitations inherent in the use of human samples (e.g. myocardium) are being overcome, at least in part, through the use of improved animal and cellular models that more closely resemble human physiology, including pigs and other large animals, and patient-derived iP cells. There is no doubt that the discovery of variants that underlie Mendelian and complex traits will lead to a much deeper understanding of disease mechanisms, and these insights should, in turn, facilitate the development of better diagnostics, prevention strategies and targeted therapeutics.

With the advent of whole-genome sequencing on a population scale,
the next challenge will be to identify rare variants with possibly larger effects. For this we will need to interpret non-coding variation in a similar way to how we predict the effect of coding variants on the protein product of genes. At present, this is not possible, but as our understanding of the regulatory genome grows, we believe this capability will be a reality in the near future. So far, the predictive value of identified variants associated with CVD is poor, and barely improves current scores based on classical risk factors (Thanassoulis et al., 2012). However, novel advances in the field, together with a better understanding of gene-environment interactions, will surely result in an increased predictive score for CVD risk (Fig. 2), and therefore a brighter outlook for early prevention and diagnosis.

The question that still lingers is how much of our susceptibility to developing CVD is written in the genome. According to a deterministic view, the probability of developing CVD and our susceptibility to environmental factors would be fully coded in the genome, and consequently our ability to prevent disease would be conditioned by the depth of our knowledge of this code. The reality is that we do not know to what extent information contained within the DNA code predicts disease, and our efforts to discover more have so far been somewhat naive. On the bright side, identifying the full predictive potential of the genome is a finite task and as such can be expected to be completed eventually, even if it takes decades or centuries. A combination of full genome sequencing, more powerful questionnaires to identify additional environmental risk factors, new animal models to study cause-effect relationships and increased knowledge of regulatory layers will provide the information needed to establish new correlations with DNA sequence. Researchers will use these blocks to build regulatory codes that will better explain the biological role of each piece of the genome so that we can use genetic variation to predict and treat disease. With the development of faster computers and increased bioinformatic expertise, we will one day be in a position to answer the question of how much information our genome really contains. Until then, we will have to take things one step at a time.

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COMPETING INTERESTS
The authors declare that they do not have any competing or financial interests.

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