Data article

Prediction and visualization data for the interpretation of sarcomeric and non-sarcomeric DNA variants found in patients with hypertrophic cardiomyopathy

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\textbf{A B S T R A C T}

Genomic technologies are redefining the understanding of genotype-phenotype relationships and over the past decade, many bioinformatics algorithms have been developed to predict functional consequences of single nucleotide variants. This article presents the data from a comprehensive computational workflow adopted to assess the biomedical impact of the DNA variants resulting from the experimental study “Molecular analysis of sarcomeric and non-sarcomeric genes in patients with hypertrophic cardiomyopathy” (Bottillo et al., 2016)\textsuperscript{1}. Several different independently methods were employed to predict the functional consequences of alleles that result in amino acid substitutions, to study the effect of some DNA variants over the splicing process and...
to investigate the impact of a sequence variant with respect to the evolutionary conservation.

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### Specifications Table

| Subject area             | Biology |
|--------------------------|---------|
| More specific subject area | In silico predictions of DNA variants |
| Type of data             | Tables, figures |
| How data was acquired    | Prediction tools: SIFT, Polyphen HDIV, Polyphen HVAR, Provean, LRT, Mutation Taster, Mutation Assessor, FATHMM, RadialSVM, LR, CADD, HSF, GERP ++, PhyloP placental, PhyloP vertebrate, SiPhy |
| Data format              | Processed, filtered and analyzed |
| Experimental factors     | Genomic DNA from peripheral blood was tested by next generation sequencing on Ion Torrent PGM (ThermoFisher, Carlsbad, CA, USA) with a custom cardiomyopathy panel |
| Experimental features    | The identified rare (Minor Allele Frequency ≤ 0.01) non-synonymous DNA changes were subjected to different in silico predictions |
| Data source location     | Rome, Italy |
| Data accessibility       | These data are with this article |

### Value of the data

- These data delineate a prompt informatic pipeline for the prioritization of the most likely pathogenic DNA variants in a clinical context.
- These data are supportive for the researchers to evaluate the prevalence of sarcomeric and non-sarcomeric gene variants in hypertrophic cardiomyopathy.
- The described computational strategy is helpful to researchers for the rapid interpretation of Variants of Unknown Significance (VUS) implicated in rare, common and complex diseases.

### 1. Data

Here we report the in silico predictions data of the non-synonymous changes found in 41 HCM patients and in 3 HCM-related cases [1] (Table 1).

### 2. Experimental design, materials and methods

#### 2.1. Analysis of the nucleotides' evolutionary conservation

Nucleotide-specific estimates of evolutionary constraint were explored by (i) GERP ++ (Genomic Evolutionary Rate Profiling); (ii) PhyloP placental; (iii) PhyloP vertebrate and (iv) SiPhy.

#### 2.2. Analysis of the splicing variants

The analysis of intronic variants leading to splicing defects was tested by Human Splicing Finder (HSF) 3.0.
Table 1

Results of the in silico predictions of the non-synonymous changes found in 41 HCM patients and in 3 HCM-related cases. Deleterious predictions are in bold.
2.3. Analysis of the missense variants

The effect of missense changes on the structure and function of a human protein was predicted by: (i) SIFT (Sorting Intolerant From Tolerant), (ii) PolyPhen-2 (Polymorphism Phenotyping v2) HDIV, that identifies human damaging mutations by assuming differences between human proteins and their closely related mammalian homologs as non-damaging; (iii) PolyPhen-2 HVAR, that identifies human disease-causing mutations by assuming common human nsSNPs as non-damaging; (iv) Provean (Protein Variation Effect Analyzer); (v) LRT (Likelihood Ratio Test) that identifies conserved amino acid positions and deleterious mutations using a comparative genomics data set of multiple vertebrate species; (vi) Mutation Taster; (vii) Mutation Assessor; (viii) FATHMM (Functional Analysis through Hidden Markov Models); (ix) RadialSVM (Radial Support Vector Machine); (x) LRT (Logistic Regression Test); (xi) CADD v1.3 (Combined Annotation–Dependent Depletion), a method for objectively integrating many diverse annotations into a single measure (C score) for each variant; and (xii) molecular modeling.

![Fig. 1. Structural comparison of wild-type and mutant forms for (a) FLH2 A37S; (b) LAMA4 E1646G; (c) MYH6 R23H; (d) MYH7 A226T; (e) MYH7 R143Q; (f) MYOM1 R711H; (g) PKP2 R767S; (h) RYR2 E1127G; (i) RYR2 R485Q. The mutation is indicated in white. The predicted structural effects of mutations are: (a, d) steric hindrance (red circles); (b) local misfolding of linker domain (orange); (c, e, f, g) loss of important inter-residues contacts; (h) loss of a \( \pi \)-anion interaction; (i) loss of a \( \pi \)-cation interaction.](image-url)
Regarding the molecular modeling, protein structure were experimentally determined by X-ray crystallography, or were inferred by homology modeling means (i.e., availability of a structural template with percentage of identity > 20%). Protein models were built using the homology modeling approach implemented in modeler-9 package [2]. PSI-BLAST was used to find suitable structural templates for each sequence to model [3]. The sequences of each protein target to model and its structural template were then aligned by using the program CLUSTALW [4] and manually manipulated to optimize the matching of several characteristics, including the observed and predicted secondary structural elements, the hydrophobic regions in the three-dimensional structures, the structurally and functionally conserved residues, and indel regions in the structures. Then, ten different models were built for each target protein and evaluated using several criteria. The model displaying the lowest objective function [5], which measures the extent of violation of constraints from the structural templates, was taken as the representative model. Superimposition and root-mean-square deviation (RMSD) calculation of Cα traces of the 10 models were performed to detect the most variable and therefore less reliable modeled regions. These invariably corresponded to loop elements.

Fig. 2. Effects of nsSNVs for: (a) the cadherin domain of DSC2. The mutant R199C in the cadherin domain of DSC2 is predicted to introduce a disulfide bond with the near Cys197 residue (Cα-Cα distance ~ 6 Å), and possibly to result in local misfolding of the cadherin domain; (b) the melibiase domain of GLA. Mutant N215S of the melibiase domain of GLA results in the loss of a glycosylated site probably affecting the protein structure and/or function; (c) the FGF13 interaction domain of SCN5. Mutation I869F localizes on a solvent-exposed hydrophobic path of the domain of interaction with fibroblast growth factor 13 (FGF13). The I869F mutation could affect the recognition of the FGF13 protein; (d) the Na-Channel of SCN5. The mutant D872N results in the loss of a negative charge that is approximately located at the Na-channel domain of SCN5, probably affecting cations conductance of the channel. The approximate position of the negatively charged Asp872 residue is shown in red, in each of the four protein subunits forming the channel.
Procheck [6] was used to monitor the stereochemical quality of the representative models, whereas Prosali [7] was used to measure the overall protein quality in packing and solvent exposure. Mutations on protein structures was carried out using the “Mutate model” script implemented in modeler-9 package [2]. The script takes as input a given three-dimensional structure of a protein (experimentally determined or predicted), and mutates a single residue. The residue sidechain's position is then optimized by energy minimization and refined by molecular dynamics simulations. Prediction of protein stability upon mutation was carried out using the DUET server [8]. Sequence identity between the modeled domain and its closest template ranged from 23% (Laminin G-like domain of LAMA4), to nearly 95% (N-terminal globular head domain of VCL). However, in spite of the low value of sequence identity measured in some cases, all of the models resulted in a good overall quality (Prosa Z-score < −2.00), except for CALR3 and SCN5. Given the short length of the predicted PB035848 domain of CALR3 (residues 294–347) and its sequence identity with its template (61%), the measured Prosa Z-score (−1.93) nonetheless indicated a model of quality comparable to a Nuclear Magnetic Resonance (NMR) structure [7] (Figs. 1 and 2).

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Appendix A. Supporting information

Supplementary data associated with this article can be found in the online version at http://dx.doi.org/10.1016/j.dib.2016.03.004.

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