VCF/Plotein: A web application to facilitate the clinical interpretation of genetic and genomic variants from exome sequencing projects

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CONFLICT OF INTEREST

No conflicts of interest.

ABSTRACT

Purpose

To create a user-friendly web application that allows researchers, medical professionals and patients to easily and securely view, filter and interact with human exome sequencing data in the Variant Call Format (VCF).

Methods

We have created VCF/Plotein, a web application written entirely in JavaScript using the Vue.js framework, available at <http://vcfplotein.liigh.unam.mx>. After a VCF is loaded, gene and variant information is extracted from Ensembl, and cross-referencing with external databases is performed via the Elasticsearch search engine. Support for application-based gene and variant filtering has also been implemented. Interactive graphs are created using the D3.js library. All data processing is done locally in the user's CPU to ensure the security of patient data.

Results

VCF/Plotein allows users to interactively view and filter VCF files without needing any bioinformatics knowledge. A number of features make it especially suited for the medical community, such as its speed, security, the ability to filter by disease or gene function, and the ease with which information may be shared with collaborators/co-workers.

Conclusion

VCF/Plotein is a novel web application that allows users to easily and interactively filter and display exome sequencing information, and that is especially suited for bench researchers, medical professionals and patients.

KEY WORDS: Exome sequencing, variant prioritisation, data visualisation, VCF, pathogenic variant.

INTRODUCTION

Exome sequencing (ES) has been highly successful at identifying genetic variation contributing to a large number of human phenotypes, from germline changes that underlie rare Mendelian disorders and complex diseases, to somatic mutations that drive carcinogenesis.^{1,2} However, the actual process of identifying disease-causing variants and mutations remains a challenging task, and often one that requires at least some bioinformatics knowledge. This is due mainly to the sheer number of variants routinely identified in ES projects, the diversity of biological mechanisms by which variants may act, and the need to integrate large amounts of information from both pathogenicity scoring algorithms and clinical and population databases.

In this context, user-friendly graphical and interactive software tools have been developed that are able to filter, display and contextualise exome sequencing data in order to accelerate the discovery of disease-causing variants. These resources vary in the amount of public information they integrate, their interactivity, and the level of bioinformatics expertise required to execute them. For example, Genome Mining (GEMINI)³ allows the user to interactively explore their own variation files and overlaps information from dbSNP,⁴ ENCODE,⁵ ClinVar⁶ and KEGG,⁷ but requires users to have a good understanding of the command line and to be able to construct MySQL queries. Similarly, VCF-Miner⁸ and BrowseVCF⁹ allow the user to interactively filter their own variant call format (VCF) annotations through a web interface, but do not leverage information from other sources such as dbSNP and COSMIC.¹⁰ Other tools like BiERapp¹¹ and exomeSuite¹² allow extensive variant filtering but do not support data visualisation. Similarly, tools that focus on variant visualisation at the protein level rather than filtering have also been developed, such as ProteinPaint,¹³ VizGVar¹⁴ and vcf.iobio.¹⁵ These resources are highly interactive, but either only display information already existing in databases without letting the user analyse their own variants (VizGVar: Ensembl),¹⁶ do not allow the user to see gene-level

information (vcf.iobio) or require the user to perform several bioinformatics steps to see their data (ProteinPaint).

Here, we introduce VCF/Plotein, a user-friendly graphical web application to both visualise and filter variant information from exome sequencing studies that requires no bioinformatics skills or knowledge. As such, this application can be used by patients to explore their own genetic information, by biologists whose projects involve exome sequencing or by medical professionals studying a particular disease. VCF/Plotein allows the user to easily load a variant call format (VCF) file, identify genes with variants, and filter and visualise the variants in any gene with information about transcripts, protein domains, variant consequences and allelic frequencies in external databases. Furthermore, this application is especially suited for the medical community mindful of patient privacy, as it allows for sharing of selected data among collaborators while avoiding having to upload sample information to the server, instead running all operations locally. The resulting protein-level graphs are fully customisable, allowing the user to effortlessly generate professional vector images from their datasets for sharing, for presentations or for publication.

METHODS

VCF/Plotein has been implemented entirely as a single-page application hosted on the Surge content-delivery network (CDN), to optimise static content downloading times across the globe. The application has been written mainly in JavaScript and uses the Vue.js-based Nuxt.js framework to control the storage, flow and presentation of information in the browser. A purpose-made API has been developed to obtain information from locally-installed external databases (gnomAD,¹⁷ dbSNP, COSMIC, ClinVar and GO term information for each annotated gene), and is running on a server with a 2-core Intel Xeon E5-4627 v4 2.60Ghz processor running a VMware 6.5.0 virtual machine over a Linux Centos 7.5 operating system. The server

also has 4GB of RAM and a solid-state hard disk drive with 1TB of storage space. VCF/Plotein works with files in the VCF format, the standard format for genetic variant data storing, which consists of fields with the genomic position, reference and alternative base changes, ID, quality score, quality filters, other metrics including custom annotations and sample genotypes.¹⁸ Upon loading, a VCF is validated and its chromosome, position, reference and alternative fields are parsed, or an error is returned to the user. After identifying the assembly version from the appropriate line in the VCF, genes with variants are quickly found by matching an interval tree algorithm to the internal coordinate indexes containing each gene's genomic positions. This generates a list with all the genes represented in the VCF, which can be filtered in different ways, such as by chromosome or by gene ontology [GO] term, in order to facilitate gene prioritisation. Once a gene is selected, information about protein-coding transcripts and functional domains is extracted from Ensembl via their REST API, independently of the existing custom annotations but without altering the original file. Dimensioning to a single gene is a key aspect for the performance of this application, since it considerably reduces computational load and information transfer over the internet. Consequences from all variants falling within the selected gene, as well as their pathogenicity scores by SIFT¹⁹ and PolyPhen,^{20,21} are obtained via the Ensembl Variant Effect Predictor.²² Cross-referencing with supported external databases is then performed via querying our internal database using the Elasticsearch search engine, which allows us to perform a smaller number of queries in an efficient manner. A flowchart depicting the process flow and system architecture of VCF/Plotein is shown in **Figure 1**. All collected information is stored as an array of objects in JSON format, returned to the web browser and depicted over a customisable plot of the primary structure of the canonical transcript (**Figure 2**). The protein graph is made using the D3.js library since it allows to easily map the information to vector graphic elements in the web page. All operations, except for the search of naked genomic positions in supported external databases, is performed locally using the user's CPU.

A menu, shown at the left-hand side of the browser, has been implemented to allow data filtering and graph customisation. The different features in this menu (Filters for transcript, functional domains and other protein features, variant type, variant frequency in external databases, variant pathogenicity score, variant clinical score and sample ID) all access information in the stored JSON object, which provides full interactivity due to the use of the Vue.js framework. A method has also been implemented that allows to export the current selected settings for any number of genes to a text file. This file can be used as a bookmark, which can be easily shared and loaded into VCF/Plotein for visualisation.

Functions to visualise and export the variant information in the customised plot to a CSV file, as well as to download and print the customised graph into a SVG or a PNG file, have also been implemented and are shown in a menu at the top right of the page.

RESULTS

Features

Overview. The only requirements to run VCF/Plotein are a computer with an internet connection and a VCF file. Once the user loads the VCF file, the genome assembly is identified, genes with variants are found, and a list of criteria is displayed to aid with gene prioritisation (by chromosome or by GO term category) (**Figure 2**, top panel). Once a gene is selected, a new page is shown with the primary protein structure of its canonical transcript with its domains and other features along with all its recorded variants. Variants are shown with an indication of their frequency among samples in the VCF, their transcript consequences, and their presence or absence in the gnomAD, dbSNP, ClinVar and COSMIC databases. The user can click on any variant to access further information about it, such as its genomic coordinates, a prediction of

their pathogenicity according to SIFT and PolyPhen and a list of carrier samples (**Figure 2**, bottom panel). The left-hand menu allows the user to load a new VCF file, to select a different gene, to select a different transcript, to select which protein domains and features to show, to filter variants, to analyse sample IDs, and finally to bookmark the selected features. Using the top menu, variant information can also be displayed and downloaded in table format, as well as printed in the SVG vector image file format or the PNG raster graphics format.

Data security. The API and the internal databases have been installed behind a Fortinet firewall, and run over an HTTPS port with a SSL certificate for secure data transfer. No sensitive sample information is uploaded to the server, as queries to the internal databases consist only of naked genomic positions. Therefore, the server does not hold or save any sample information, an important feature given the data security policy that many patient-focused sequencing projects are bound by. All data processing, including construction of the JSON object and graphing of protein structures, is done locally.

Variant filtering and visualisation. Variants falling in any selected protein-coding transcript from any gene can be filtered and plotted. Users can filter variants by protein consequence (for example, to show only stop gains or splice site mutations), by clinical prediction (for example, to display only those present in the ClinVar or COSMIC datasets), by pathogenicity score (for example, only those scored as deleterious by SIFT and/or PolyPhen) or by their presence in allele frequency databases (for example, to display only variants not previously seen in gnomAD or dbSNP). Users can also select which protein domains and features to display (from InterProScan²³ and Pfam²⁴ and the ncoils,²⁵ SEG,²⁶ SignalP²⁷ and TMHMM²⁸ programs, as annotated by Ensembl), as well as customise the colours for each of these features. The customised protein plot can then be exported as an SVG or PNG file to use in presentations or

publications. Because of the nature of this web application, it is also well-suited to visualise information from large databases such as dbSNP or ClinVar.

Performance. VCF/Plotein is able to process VCF files from exome sequencing studies in a reduced time frame. We have tested our application with different file sizes in a number of system architectures and web browsers. The results are shown in **Table 1**.

Bookmarks. Bookmarks allow users to easily save any selected features from any number of genes in a text file (in JSON format) which can be easily shared so others can load it into VCF/Plotein and view and interact with this information. As only the information regarding the selected gene transcripts is saved, this function can be used by teams of clinicians and researchers studying a few candidate genes from exome-wide sequencing studies in order to facilitate data transfer and interpretation.

Comparison with other similar tools. VCF/Plotein combines variant filtering capabilities with an intuitive and interactive platform to visualise and customise protein plots. Furthermore, it does not upload any sensitive sample information to the server, instead running all operations locally. For these reasons, and because it requires no bioinformatics knowledge, we believe it is well-suited for patients exploring their own genetic data, for biologists analysing exome sequencing data, and for teams of scientists and medical professionals studying patients with a particular disease. Other available tools perform some of these functions, but either require at least some bioinformatics expertise, do not leverage information from external databases, or do not allow users to visualise their own exome data (**Table 2**).

DISCUSSION

The ability to easily filter and display genomic information at the protein level is expected to contribute importantly to the identification of genetic variants conferring high-risk to develop various diseases by allowing non-bioinformaticians (e.g. many medical doctors and bench researchers) to perform variant prioritisation. Here, we have described VCF/Plotein, a novel web application capable of loading VCF files from exome sequencing projects in order to display, in a graphical and interactive manner, variants falling in any chosen transcript from any gene. Furthermore, variants can be filtered and customised to display only those of interest to the user, which can then be easily shared via a simple text file that can also be loaded into the web application. This kind of analyses can sometimes take an experimented bioinformatician hours or days (given the need to download databases, create custom scripts for filtering and find adequate tools for data visualisation), however here we have simplified this process and shortened the time required for analysis to minutes or seconds.

Our methodology also entails a number of advantages: By building VCF/Plotein as a single-page application, we are able to implement an iterative and incremental software development strategy, undertake constant improvements and regularly add new features. Furthermore, it allows us to easily keep our internal data structures up-to-date, importing the latest versions of external databases as they are released.

We anticipate that VCF/Plotein will allow researchers, especially in small labs, to focus on biology-relevant questions instead of having to learn to install software dependencies, learn to use variant-annotation and cross-referencing tools, and become familiar with the UNIX and/or MySQL command line. As direct-to-consumer genetic testing becomes even more accessible, we also believe that this application can be used by patients analysing their own genetic information. By combining variant filtering and annotation in a single graphical and interactive tool, we hope that variant prioritisation will become easier, faster and more intuitive.

AVAILABILITY

VCF/Plotein is freely available at <http://vcfplotein.liigh.unam.mx>

SUPPLEMENTARY DATA

No supplementary data

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TABLE AND FIGURES LEGENDS

Figure 1. Chart depicting the VCF/Plotein process flow and system architecture. The colours of each box in the top panel correspond to the colour of the system architecture element in which it is executed (in the bottom panel), *e.g.*, actions in green boxes are executed by the user, those in blue boxes are executed by the user's browser, and actions in red, purple and orange are processes used to obtain information from diverse databases.

Figure 2. VCF/Plotein working screens. Top panel: Loading and data filtering screen. In the left column, the loaded VCF or bookmark is shown along the identified genome assembly. In the middle column, different gene filtering criteria are shown. The right column allows the user to select an individual gene to examine in detail. Bottom panel: Screen depicting protein-level information for the selected gene. The primary protein structure for the canonical transcript is shown alongside protein domains and other features, and the genetic variants falling on it are shown as lollipops. The colours of these indicate the protein consequence, and the squares below indicate the presence or absence of the variant in different external databases. Upon selection of a particular variant, other relevant data are shown in information bars at the top and bottom. Left-hand side and top menus allow the user to customise and interact with the plot. The depicted VCF contains variants published elsewhere (Ref. ²⁹).

Table 1. Performance of VCF/Plotein after loading three different VCF files in different operating systems with a range of hardware specifications. Tests were performed in the Google Chrome browser. All times are in milliseconds (ms).

Table 2. Comparison of the main features of VCF/Plotein with those of other similar tools.

Figure 1: Process flow

System architecture

Open VCF or Bookmark

BAP1_OShea_etal.vcf

Selected file:

BAP1_OShea_etal.vcf

Reference genome: (38)

Human assembly GRCh38

Filter genes

Chromosome

Biological
process

Molecular
function

Cellular
component

BIOLOGICAL PROCESS

Search a Go-term

regulation of cell growth

cellular protein modification process

ubiquitin-dependent protein catabolic process

negative regulation of cell proliferation

response to inorganic substance

protein deubiquitination

Genes (1)

Filtered by goterm (GO:0016579)

Search a gene	
NAME	ID
BAP1	ENSG00000163930

GENERAL		VARIANTS			SAMPLES
GENE NAME	ENSEMBL ID	CHROMOSOME	GENOMIC POSITION	PROTEIN POSITION	SAMPLES
BAP1	ENSG00000163930	3	52403593	518	SAMPLES IN VARIANT: 1
TRANSCRIPT ID	ENST00000460680	REFERENCE	ALTERNATIVE	AMINOACID CHANGE	SC_MC5380447
		G	A	R/W	
gnomAD					
dbSNP					
ClinVar					
COSMIC					
PREDICTION		CLINICAL		POPULATION	
SIFT SCORE	POLYPHEN SCORE	ClinVar	COSMIC	gnomAD	dbSNP
0.1	0.993	Present	Present	Absent	Present
SIFT PREDICTION	POLYPHEN PREDICTION	CLINICAL SIGNIFICANCE		TOTAL POPULATION	
Tolerated	Probably Damaging	Susceptibility to melanoma		0.1	

	Computer specifications	Processes	Single gene VCF Size: 2.9 mb Number of genes: 1 Gene selected: 490 variants Samples: 201	ClinVar VCF Size: 170.7 mb Number of genes: 7357 Gene selected: 340 variants Samples: NA	COSMIC VCF Size: 421.8 mb Number of genes: 23141 Gene selected: 1979 variants Samples: NA
1	Operating system: macOS Mojave Processor: 2.2 GHz intel core i7 RAM: 8 GB 1600MHz DDR3 Hard disk: 500 GB	File opening and parsing Generation of list of genes Creation of GO term filters Fetching of variant information	62 80 5 5724	1665 2821 1486 6499	4390 32300 11277 22192
2	Operating system: Manjaro Linux Processor: 1.7GHz AMD A8 x 4 RAM: 7 GiB Hard disk: 750 GB	File opening and parsing Generation of list of genes Creation of GO term filters Fetching of variant information	222 16 255 6159	11528 4954 12566 31141	60492 75800 32418 106972
3	Operating system: Windows 10 Processor: intel core i7-6500U RAM: 8 GB 1600MHz DDR3 Hard disk: 1 TB	File opening and parsing Generation of list of genes Creation of GO term filters Fetching of variant information	18 32 5 4506	1033 1159 909 6754	10599 78363 10515 25364
4	Operating system: macOS Mojave Processor: 2.2 GHz intel core i7 RAM: 8 GB 1600MHz DDR3 Hard disk: 500 GB	File opening and parsing Generation of list of genes Creation of GO term filters Fetching of variant information	35 15 112 4974	2666 10648 10488 7042	33034 110945 29218 19251

	VCF/Plotein	GEMINI	VCF-miner	BrowseVCf	BiERapp	exomeSuite	PeCan	VCF iobio	VizGVar
Does not require installation	✓	x	x	x	x	x	✓	✓	✓
Does not require command line knowledge	✓	x	x	x	✓	x	✓	✓	✓
Graphical user interface	✓	✓	✓	✓	✓	✓	✓	✓	✓
No pre-processing steps	✓	x	x	x	✓	x	x	✓	✓
Database cross-referencing	✓	✓	x	x	✓	✓	✓	x	✓
Protein-level information plot	✓	x	x	✓	x	x	✓	x	✓
Local sample processing (security)	✓	✓	✓	✓	✓	✓	x	x	NA
Export as SVG	✓	NA	NA	NA	x	x	✓	x	✓