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RESEARCH ARTICLE

BLAT2DOLite: An Online System for Identifying Significant Relationships between Genetic Sequences and Diseases

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Abstract

The significantly related diseases of sequences could play an important role in understanding the functions of these sequences. In this paper, we introduced BLAT2DOLite, an online system for annotating human genes and diseases and identifying the significant relationships between sequences and diseases. Currently, BLAT2DOLite integrates Entrez Gene database and Disease Ontology Lite (DOLite), which contain loci of gene and relationships between genes and diseases. It utilizes hypergeometric test to calculate P-values between genes and diseases of DOLite. The system can be accessed from: <u>http://123.59.132</u>. <u>21:8080/BLAT2DOLite/BLAT2DOLite/BLAT2DOLiteIDMappingPort?wsdl</u>.

Introduction

Identifying significantly related diseases of genes has drawn more and more attention in interpreting molecular functions [1-13]. For example, through exploiting the significant relationships between diseases and altered genes by promyelocytic leukemia protein (PML) based on microarray analysis, Anida et al. identified the role of PML in diseases other than cancers [1]. Jiny et al. exploited overlapping between disease-related genes and inflammatory genes to explore core transcriptional regulators of inflammatory genes in coronary artery disease [2].

Enrichment analysis is an effective method to identify the significant relationships between diseases and genes. To this end, a disease vocabulary and a data set of associations between diseases and genes are needed first. Many databases are suitable for this purpose, in which Online Mendelian Inheritance in Man (OMIM) [14] and Gene References Into Function (GeneRIF) [15] have been most commonly used. OMIM is a database that concerns genetic disorders and its induced genes. In contrast, GeneRIF is more comprehensive, which is initiated by the National Library of Medicine (NLM) to link published data to Entrez Gene entries. GeneRIF consists of an Entrez Gene ID, a short text (under 255 characters), and the PubMed identifier (PMID) of the publication that provides evidence for the assertion in that text. Then, gene-disease relationships from the GeneRIF database were discovered [16] by Unified Medical Language System (UMLS) [17] MetaMap Transfer tool (MMTx) [18]. Here, disease terms were filtered by Disease Ontology (DO) [19]. In consideration that a simplified version of vocabulary could be helpful for integrating overview of molecular and cellular biology by combining and removing fine-grained terms [20,21], a simplified vocabulary list from the DO called Disease Ontology Lite (DOLite) [22] was constructed for enrichment analysis.

Many tools have been developed for the ease of accessing the significant relationships between diseases and genes, such as DAVID [23], FunDO [22], DOSE [24], DOSim [25], and GeneAnswer [26]. DAVID was an early bioinformatics analytic tool for systematically extracting biological meaning from large gene/protein lists. In contract, FunDO, DOSE, DOSim, and GeneAnswer can be used to study the significant relationship between diseases and genes. Though gene symbols or gene IDs can be analysed by existing tools, sequence data cannot be processed by all of these five tools. With the development of the next-generation sequencing technology, a large number of sequence data have been produced. Meanwhile, sequence alignment tools have been developed to identify the loci of sequence [27,28]. Therefore, analysing the relationship between sequence data and diseases is a critical challenge.

In this paper, we presented an online tool BALT2DOLite to annotate human genes and diseases, and to identify the significantly related diseases of sequences. Through BLAT2DOLite, sequences were first mapped to their locus by BLAT, and then these sequences were mapped to genes. According to associations between diseases of DOLite and genes, hypergeometric test was exploited to calculate the significant relationships between them. The system can be accessed from: <u>http://123.59.132.21:8080/BLAT2DOLite</u>. For easing to invoke the functions of BLAT2DOLite locally, a web service was also provided, which is described in: <u>http://123.59.</u> <u>132.21:8080/BLAT2DOLite/BLAT2DOLiteIDMappingPort?wsdl</u>.

Materials and Methods

Data Collection

Data sets of BLAT2DOLite were from open source databases. All of these databases were listed in the <u>Table 1</u>. For example, disease terms and relationships between these diseases and genes were from DOLite [22]. Currently, DOLite contains 15,016 associations between 560 diseases and 3,966 genes. In addition, a human reference genome (hg19) [29] was originated from UCSC Genome Browser [30]. In order to retrieve mappings from locus to genes, Entrez Gene database [31] was integrated in our system.

Table 1. Data source	es and tools used for identifying significant relationships between sequences and
diseases.	

Data source / tool	Web site (Date of download)
DO	https://diseaseontology.svn.sourceforge.net/svnroot/diseaseontology/trunk/ (Jan 2016)
DOLite	http://fundo.nubic.northwestern.edu/ (Jan 2016)
Entrez Gene database	http://www.ncbi.nlm.nih.gov/gene (Jan 2016)
hg19	ftp://hgdownload.cse.ucsc.edu/gbdb/hg19/ (Jan 2016)
BLAT	http://genome.ucsc.edu/cgi-bin/hgBlat (Jan 2016)

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The Process of BLAT2DOLite

According to our system, significantly related diseases of sequences could be identified, the process of which was described in the Fig 1 as following.

Step 1: Mapping sequence to locus. Sequences could be mapped to a human reference genome (hg19) by BLAT, which is an open source software for finding loci of sequences. After mapping by BLAT [32], the location with the longest sequence mapping is selected.

Step 2: Annotating locus, gene symbol, or gene ID with diseases. Sequences in the previous step could be related to genes based on their locus. Here, two types of relevance were used for annotation: 1) Contain: the loci of gene is in the locus of sequences or the locus of sequences is in the loci of gene; 2) Intersect: The loci of gene covers the locus of sequences partly. Then, based on the relationships between genes and diseases of DOLite, sequences could be annotated with human diseases.

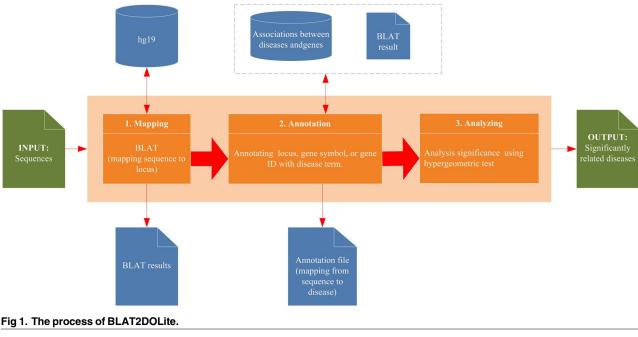
Method for analyzing the significant relationship between sequences and diseases. Here, hypergeometric test was utilized for analyzing the significant relationship between sequences and diseases. The formula for calculating P-value is as follows:

$$p - value = 1 - \sum_{0 \le i \le x} \frac{C_M^i \times C_{N-M}^{k-i}}{C_N^k}.$$
(1)

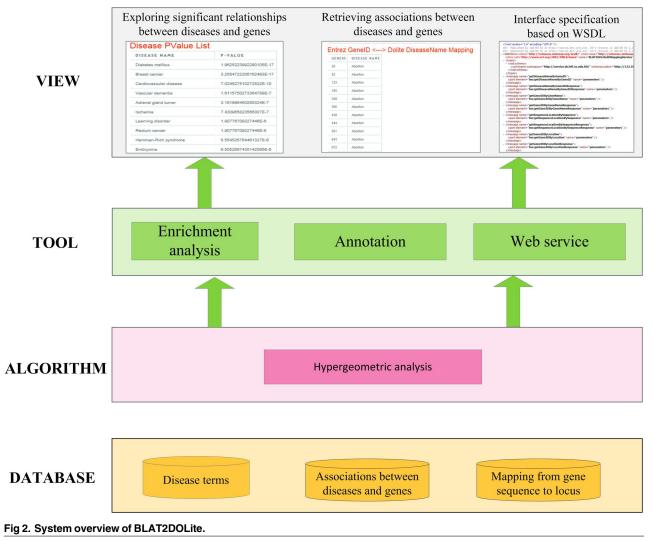
Taking breast cancer as an example, N indicates the number of genes related by all of diseases, M indicates the number of genes related with breast cancer, k indicates the number of genes related with sequences, x indicates the number of common genes related with sequences and breast cancer.

Implementation

BLAT2DOLite has been implemented on a JavaEE framework and run on the web server (2-core (2.26 GHz) processors) of UCloud [33]. The four-layer architecture involving



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DATABASE, ALGORITHM, TOOLS, and VIEW layer is shown in the Fig 2. The detailed description of the architecture is as following.

- 1. DATABASE layer. This layer stores locus of genes, disease terms and associations between human genes and diseases. These data are used by ALGORITHM layer and TOOL layer for annotating human genes and diseases and identifying the significant relationships between human diseases and sequences, respectively.
- 2. ALGORITHM layer. Hypergeometric analysis is implemented for calculating the significant relationships between diseases and sequences.
- 3. TOOL layer. The system provides two types of functions including annotating human genes and diseases and identifying the significant relationships between sequences and diseases. Furthermore, the functions of this system can be accessed based on our web service [34].
- 4. VIEW layer. Webpages are provided for viewing all the results based on TOOL layer. For example, the relationship between human diseases and genes can be shown, and the

significant relationship between sequences and diseases can also be obtained. In addition, the interface specification of our web service can be accessed from the web.

Results

The system could be used for annotating human genes and diseases, and identifying the significant relationships between sequences and diseases. The details about the access to these two functions are described as follows.

A case for annotating human genes and diseases

Human genes and diseases can be annotated from the web (<u>http://123.59.132.21:8080/</u> <u>BLAT2DOLite/geneid2diseasename.jsp</u>), a case of which is shown in Fig 3.

According to the figure, the system could return diseases after submitting an Entrez Gene ID. In this case, the inputted gene ID was '28'. And diseases could be affected by this gene were listed in the result page, such as bladder cancer, squamous cell cancer, and so on. Similarly, the system could return Entrez Gene IDs after submitting a disease term. In this case, the inputted disease term was 'Abortion'. And gene IDs could induce this disease were listed in the result page, such as '52', '153, and so on.

A case for identifying the significant relationships between sequences and diseases

The significantly related diseases of sequences could be identified from the web (<u>http://123.59.132.21:8080/BLAT2DOLite/sequence.jsp</u>), a case of which is shown in Fig 4.

	Entrez		Dolite DiseaseName Mapping
	GENEID	DISEASE NAME	Donte Diseaservarite Mapping
	28	Abortion	
	28	Bladder cancer	
	28	Infection	
ID Mapping for GeneID to DiseaseName	28	Intestinal disease	
	28	Leukemia	
Submit a Entrez gene ID, then return gene-related disease names of DOLite. Input Entrez GeneID	28	Squamous cell cancer	
(Disease Related) as follows.	28	Yersinia infection	
28 submit			
Submit a disease name of DOLite, then return disease-related Entrez gene IDs. Input DOLite term as			
Submit a disease name of DOLite, then return disease-related Entrez gene IDs. Input DOLite term as follows.	Entrez	GeneID <>	Dolite DiseaseName Mapping
follows.	Entrez GENEID	GeneID <> disease name	Dolite DiseaseName Mapping
			Dolite DiseaseName Mapping
follows.	GENEID	DISEASE NAME	Dolite DiseaseName Mapping
follows.	GENEID 28	Abortion	Dolite DiseaseName Mapping
follows.	GENEID 28 52	DISEASE NAME Abortion Abortion	Dolite DiseaseName Mapping
follows.	GENEID 28 52 123 185	DISEASE NAME Abortion Abortion Abortion	Dolite DiseaseName Mapping
follows.	GENEID 28 52 123 185 350	DISEASE NAME Abortion Abortion Abortion Abortion Abortion	Dolite DiseaseName Mapping
follows.	GENEID 28 52 123 185 350 356	DISEASE NAME Abortion Abortion Abortion Abortion Abortion	Dolite DiseaseName Mapping
follows.	GENEID 28 52 123 185 350 356 430	DISEASE NAME Abortion Abortion Abortion Abortion Abortion Abortion	Dolite DiseaseName Mapping
follows.	GENEID 28 52 123 185 350 356 430 444	DISEASE NAME Abortion Abortion Abortion Abortion Abortion Abortion Abortion	Dolite DiseaseName Mapping
follows.	GENEID 28 52 123 185 350 356 430	DISEASE NAME Abortion Abortion Abortion Abortion Abortion Abortion	Dolite DiseaseName Mapping
follows.	GENEID 28 52 123 185 350 356 430 444	DISEASE NAME Abortion Abortion Abortion Abortion Abortion Abortion Abortion	Dolite DiseaseName Mapping
follows.	GENEID 28 52 123 185 350 356 430 444 581	DISEASE NAME Abortion Abortion Abortion Abortion Abortion Abortion Abortion Abortion	Dolite DiseaseName Mapping

Fig 3. Schematic workflow of annotating human genes and diseases.

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						lence			
E:\1.fa				1		br	owse submit		
sequence>	Locus M	apping	g 🚽				Disease DV(slug)		
GI			OSOME	OSOME START POSITION		END POSITION	Disease PValue List		
gij224589803:689863	8-6929976	12		6898637		6929976	DISEASE NAME	P-VALUE	
gij224589818:c32152099-32148745 6		6		32148744		32152099	Breast cancer	2.5884948043369204E-1	
gi[224589815:148415658-148460790 3		3		148415657		148456156	Diabetes mellitus	1.6040063901334118E-1	
gij224589822:115301	958-115306225	х		115301957		115306225	Cardiovascular disease	2.198114934301251E-7	
gij224589810:4540903	39-45412650	19		45409038		45412650	Embryoma	6.335887514922207E-7	
gij224589800:c150849244-150782181 1		1		150782180		150822426	Ischemia	1.312599780431108E-6	
gij224589800:c226595801-226548392 1		1		226548	391	226588891	Uterine fibroids	1.94171443244803E-6	
gij224589818:7727011-7881961		6	7727		0	7767256	Vascular dementia	2.9065064906677682E-6	
gij224589816:c75719882-75671448		4	4 756		47	75711946	Adrenal gland tumor	5.685453939439488E-6	
gi224589804:73633142-73651676		13			41	73651676	Learning disorder	1.1573787025989483E-5	
		11	344604		71	34493607	Malignant fibroxanthoma	1.1573787025989483E-5	
-							Rectum cancer	1.1573787025989483E-5	
_ocus>Ge	nelD Map	ping					Macular degeneration	5.951551923522036E-5	
CHROMOSOME	E START POSTITION		N END POSTION		GENEID	LOCUS GENE RELATION	Lupus erythematosus	6.174190084452455E-5	
12	6898637		6929976		920	contain	Hamman-Rich syndrome	9.747934388488295E-5	
6	32148744		32152099	52099 177 0		contain	Biliary cancer	1.1503859420712234E-4	
6	32148744		32152099	9 5089 8		adjacent	Brain ischemia	1.1503859420712234E-4	
6	32148744		32152099	6048 8		adjacent	Mycoses	1.1503859420712234E-4	
3	148415657		14845615	156 185 i		intersect	Rheumatoid arthritis	1.5665949118280772E-4	
Х	115301957		11530622	186		contain	Hypoglycemia	1.7220970811747047E-4	
19	45409038	038		348		contain	Kidney tubular necrosis, acute	1.7220970811747047E-4	
1	150782180		150822426		405	intersect	IGA glomerulonephritis	1.9653168802555299E-4	
1	150782180		150822426		1513	adjacent	AIDS	2.4060714985050038E-4	
1	226548391		22658889	226588891 142		intersect	Chondrosarcoma	2.4060714985050038E-4	
6	7727010		7767256		654	intersect	Drug-Induced dyskinesia	2.4060714985050038E-4	
4	75671447		75711946 685		685	intersect	Werner syndrome	2.4060714985050038E-4	
13	73633141		73651676		688	contain	Squamous cell cancer	2.6795935726662244E-4	
11	34460471		34493607 847					3.201622949902602E-4	

Fig 4. Schematic workflow of identifying significant relationships between sequences and diseases.

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In this system, DNA sequences with FASTA format, in which nucleotides are represented using single-letter codes, could be submitted as an input. This format originates from the FASTA software package [35], but has now become a standard in the field of bioinformatics.

According to the schematic workflow of BLAT2DOLite in the Fig_1. First, sequences could be mapped to locus in the hg19. This mapping result could be returned to the result page. Next, the locus of these sequences could be mapped to Entrez Gene IDs based on the integrated Entrez Gene database. The corresponding associations between locus of these sequences and the locus of genes could also be shown in the result page. Then, these mapped gene IDs were annotated with diseases by BLAT2DOLite. The annotation result was not shown in this result page, in case the annotation function was provided by the system in the annotation page. Finally, the hypergeometric test was used to calculate P-values between these mapped genes and each disease of DOLite. Diseases with P-value less than 0.05 could be shown in the result page.

In the case shown in the Fig 4, the sequences in the web page were used as input. And the result page including 'Sequence-Locus Mapping', 'Locus-Gene ID Mapping' and 'Disease P-value' sections could be returned. In the 'Sequence-Locus Mapping' section, the identifiers of mapped sequences were shown in the first column of the table. And the mapped chromosome, start position, and end position of sequences in the same line were listed in the next three columns, respectively. For example, sequences gi|224589803:6898638–6929976 were mapped to locus from 6898637 to 6929976 in the twelfth chromosome. In the 'Locus-Gene ID Mapping' section, the relationships between loci of sequences and Entrez Gene IDs could be obtained. For example, in the first line of the result table of this section, the loci of gi|224589803:6898638–6929976 was mapped to Entrez Gene '920'. In the 'Disease P-value' section, significantly related diseases of these sequences were listed ranked by the P-values in descending order. In this case, diabetes mellitus was identified as the most significant disease of these sequences, so it was listed in the top of the corresponding result table.

Web service of BLAT2DOLite

All the functions of our system were implemented as a web service through the JAVA API for XML Web Services (JAX-WS). The detailed description of our web service can be accessed from the following website: <u>http://123.59.132.21:8080/BLAT2DOLite/BLAT2DOLiteIDMappingPort?</u> wsdl. According to the interface of our web service, users can easily introduce the function of BLAT2DOLite locally.

Conclusion

In this paper, an online system was presented for annotating human genes and diseases and identifying the significant relationships between sequences and diseases. For identifying the relationships between sequences and diseases, BLAT and the Entrez Gene database were integrated to map sequence to Entrez Gene ID. In this system, associations between human genes and diseases of DOLite were utilized for calculating the significant relationships between them. Furthermore, a web service was provided for the ease of introducing the function of BLAT2-DOLite locally.

Author Contributions

Conceived and designed the experiments: LC YH. Performed the experiments: LC SZ YH. Analyzed the data: LC SZ YH. Contributed reagents/materials/analysis tools: LC. Wrote the paper: LC.

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