

XXX Symposium on Bioinformatics
and Computer-Aided Drug Discovery



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Institute of Biomedical Chemistry

ANALYSIS OF CHEMICALS-VIRUS-HOST INTERACTIONS BASED ON LARGE-SCALE BIOMEDICAL TEXT AND DATA MINING

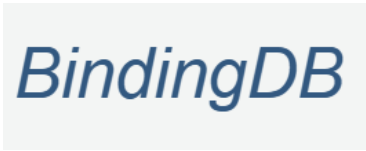
Olga Tarasova, PhD

Laboratory of Structure-Function Based Drug Design, Senior Scientist

Laboratory of Big Data Analysis for Digital Pharmacology, Head

Institute of Biomedical Chemistry, Moscow, Russia

Large-scale biomedical data are available in the databases and scientific publications



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journal homepage: www.elsevier.com/locate/cbpb

**QSAR
HIV-1 F**

**Olga A. T.
Alexey V.**
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‡CADD G
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21702, Ur

Perspectives on transcriptomics in animal physiology studies

Christopher J. Martyniuk*

Department of Physiological Sciences and Center for Environmental and Human Toxicology, University of Florida Genetics Institute, Interdisciplinary Program in Biomedical Sciences Neuroscience, College of Veterinary Medicine, University of Florida, Gainesville, FL 32611, USA

<https://pubs.acs.org/doi/10.1021/jm300131x>

ABSTRACT

Abstrac

Large-approa variety activit; reduce how be predict databa presen comm; that th compil trainin type of predict databa of ther metho of resu

Graphic

Reductionist approaches longer restricted to a sel estimate) over the past d under conditions of hypx understanding of the m exquisitely sensitive to a consider experimental e experiments in order to RNA-seq should be expl microarrays. Incorporati not be viewed as a "va experimental controls in tests (round robin exper reproducibility of trans transcriptomics and mul encouraged to consult li animals

1. Transcriptomics i RNA sequencing

Transcriptomics is a cell or tissue. The sc of experience; it per physiology and ecoph table interrogating th quencing technologie oped at academic in commercial compani continue to push bour sequencing depth is i collect gigabytes of se from insects to fish, i read sequencers (Paet identify rare alternati genetic modifications 2020; Burgess, 2020;

* Corresponding autho
E-mail address: cmart

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The Experimental Uncertainty of Heterogeneous Public K_i Data

Christian Kramer*, Tuomo Kalliokoski*, Peter Gedeck, and Anna Vulpetti

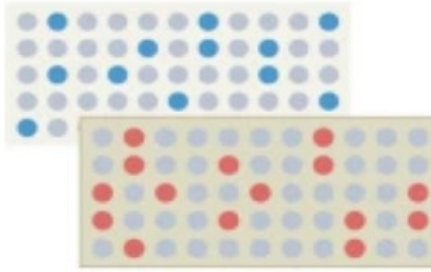
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Cite this: *J. Med. Chem.* 2012, 55, 11, 5165–5173
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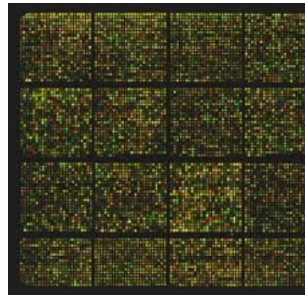
SUBJECTS: Assays, Computational chemistry, Molecular struc



Genome analysis/ genome-wide association studies



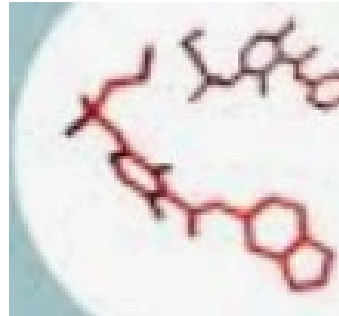
Epigenome-wide/ Methylome-wide association studies



Transcriptome data analysis/ transcriptome wide association studies



Proteome-wide association studies



Metabolome-wide association studies

MVIP: multi-omics portal of viral infection

Zhidong Tang^{1,†}, Weiliang Fan^{1,†}, Qiming Li^{1,†}, Dehe Wang^{2,†}, Miaomiao Wen², Junhao Wang¹, Xingqiao Li¹ and Yu Zhou^{1,2,3,4,*}

¹State Key Laboratory of Virology, College of Life Sciences, Wuhan University, Wuhan 430072, China, ²Institute for Advanced Studies, Wuhan University, Wuhan 430072, China, ³RNA Institute, Wuhan University, Wuhan 430072, China and ⁴Frontier Science Center for Immunology and Metabolism, Wuhan University, Wuhan 430072, China

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ABSTRACT

Virus infections are huge threats to living organisms and cause many diseases, such as COVID-19 caused by SARS-CoV-2, which has led to millions of deaths. To develop effective strategies to control viral infection, we need to understand its molecular events in host cells. Virus related functional genomic datasets are growing rapidly, however, an integrative platform for systematically investigating host responses to viruses is missing. Here, we developed a user-friendly multi-omics portal of viral infection named as MVIP (<https://mvip.whu.edu.cn>). We manually collected available high-throughput sequencing data under viral infection, and unified their detailed metadata including virus, host species, infection time, assay, and target, etc. We processed multi-layered omics data of more than 4900 viral infected samples from 77 viruses and 33 host species with standard pipelines, including RNA-seq, ChIP-seq, and CLIP-seq, etc. In addition, we integrated these genome-wide signals into customized genome browsers, and developed multiple dynamic charts to exhibit the information, such as time-course dynamic and differential gene expression profiles, alternative splicing changes and enriched GO/KEGG terms. Furthermore, we implemented several tools for efficiently mining the virus-host interactions by virus, host and genes. MVIP would help users to retrieve large-scale functional information and promote the understanding of virus-host interactions.

INTRODUCTION

Viruses are everywhere, comprising an enormous proportion of our environment, in both quantity and total mass (1). Many viral infections cause human diseases (2,3). More than 12% new cancer cases were attributable to oncoviruses, such as hepatitis B or C virus (HBV or HCV), Epstein-

Barr virus (EBV), Kaposi's sarcoma herpes virus (KSHV), and human papillomavirus (HPV) (4–6). Recently, severe acute respiratory syndrome coronavirus 2 (SARS-CoV-2) caused the COVID-19 disease, and resulted in a global pandemic and millions of deaths (7–9). Viral infections generally cause dysregulated gene expression and abnormal RNA processing (10–13). In mammals, viral infections can lead to local inflammatory responses and innate immune responses called as 'cytokine storm' (2). For example, SARS-CoV-2 broadly alters gene expression programs in human cells and disrupts splicing to suppress host defences (14,15). In addition, SARS-CoV-2 RNAs can bind and repurpose host RNA-binding proteins (RBPs), which is one of the pathogenetic factors (16–18). Moreover, viral infections can also change the epigenetic states and RNA modifications of hosts (19–22). To better understand how viruses affect hosts at molecular level, we need to integrate various types of omics data and systematically analyse the many-to-many virus-host interactions genome-wide.

In recent years, the studies of genome, structure and taxonomy have been rapidly developed for viral species, including ViPR (23), VIPERdb (24,25), IMG/VR v2.0 (26) and ICTV (27) databases. Moreover, it is found that the molecular network of host in many cancers are perturbed by viral proteins (17). Therefore, the relevant resources of biological pathway and network signatures associated with virus were developed, such as KEGG (28) and PAGER (29,30). In addition, multiple types of raw sequencing data under viral infection are deposited into the NCBI GEO and SRA (31,32) databases. These data were separately generated in different studies to uncover the cellular events in various species with different viral infections. However, an integrative multi-omics database of virus-host interactions for multiple species/viruses, enabling users to mine relevant data jointly, is missing.

Here, we have developed a user-friendly multi-omics portal of viral infections across different species, named MVIP (<https://mvip.whu.edu.cn>). We firstly manually collected available high-throughput sequencing data under viral infections, and also the description of these data (metadata). We unified detailed metadata including virus, host species,

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[†]The authors wish it to be known that, in their opinion, the first three authors should be regarded as joint First Authors.

Integration of data obtained in the specific E/G/M/T/PWAS may be helpful for the comprehensive understanding of particular diseases mechanisms, and the methods for meta-analysis were proposed and described [Nan Wang, Shuilin Jin; Methods Mol. Biol., 2022].

Text and Data Mining Tools Can Help Researchers

HIV-1 resists MxB inhibition of viral Rev protein.

1 Wang Z, Chai K, Liu Q, Yi DR, Pan Q, Huang Y, Tan J, Qiao W, Guo F, Cen S, Liang C.

Cite Emerg Microbes Infect. 2020 Dec;9(1):2030-2045. doi: 10.1080/22221751.2020.1818633.

PMID: 32873191 [Free article.](#)

Share Here, we report a new antiviral mechanism in which MxB restricts the nuclear import of **HIV-1** regulatory protein Rev, and as a result, diminishes Rev-dependent expression of **HIV-1** Gag protein. ...In addition, **HIV-1** can overcome this **inhibition** by MxB th ...

Complex genetic encoding of the hepatitis B virus on-drug persistence.

2 Thai H, Lara J, Xu X, Kitrinos K, Gaggar A, Chan HLY, Xia GL, Ganova-Raeva L, Khudyakov Y.

Cite Sci Rep. 2020 Sep 23;10(1):15574. doi: 10.1038/s41598-020-72467-9.

PMID: 32968103

Share Tenofovir disoproxil fumarate (TDF) is one of the nucleotide analogs capable of inhibiting the reverse transcriptase (RT) activity of **HIV** and hepatitis B virus (HBV). ...These pervasive mechanisms are insufficient to prevent viral **inhibition** completely but may contr ...

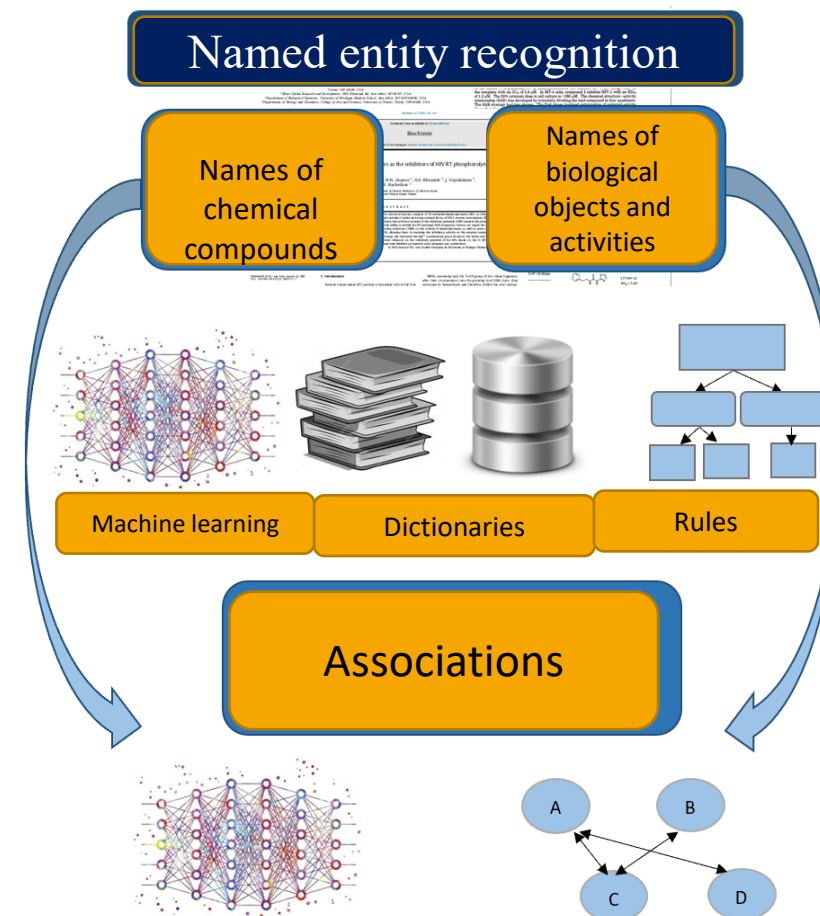
Integrase-RNA interactions underscore the critical role of integrase in HIV-1 virion morphogenesis.

3 Elliott JL, Eschbach JE, Koneru PC, Li W, Puray Chavez M, Townsend D, Lawson DQ, Engelman AN, Kvaratskhelia M, Kutluay SB.

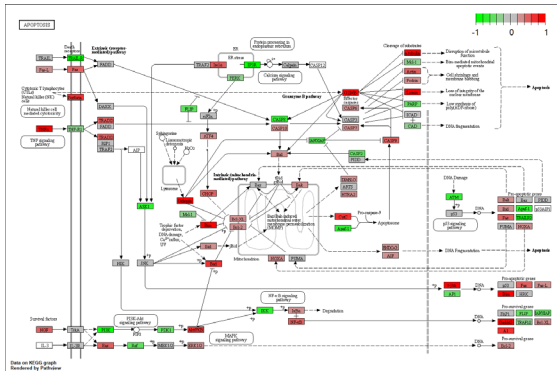
Cite Elife. 2020 Sep 22;9:e54311. doi: 10.7554/eLife.54311. Online ahead of print.

PMID: 32960169

Inhibition of IN-RNA interactions resulted in mislocalization of the viral ribonucleoprotein complexes

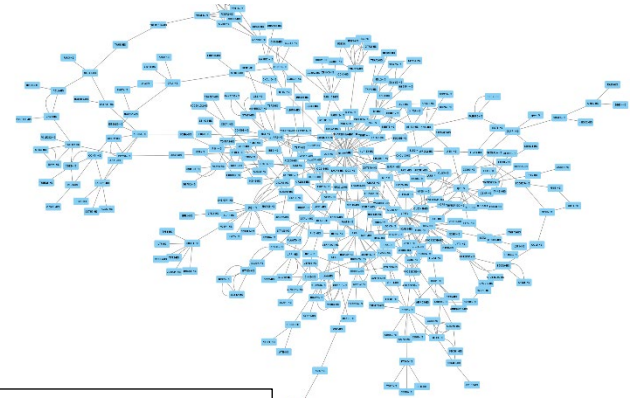
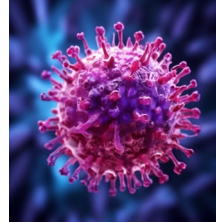


Textome - is a comprehensive set of biological literature that contains useful information and provides retrieval new knowledge using bioinformatics, ML and AI.

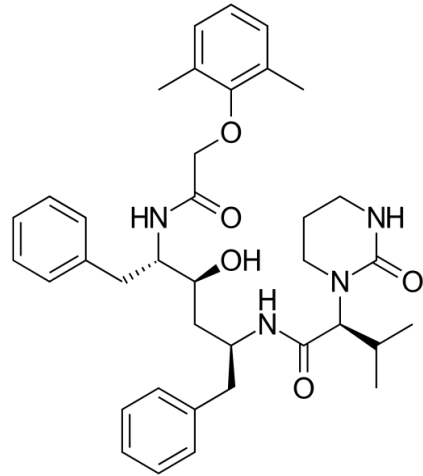


Textome
Transcriptome

An analysis of
HIV infection
progression
velocity



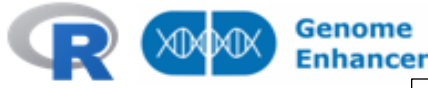
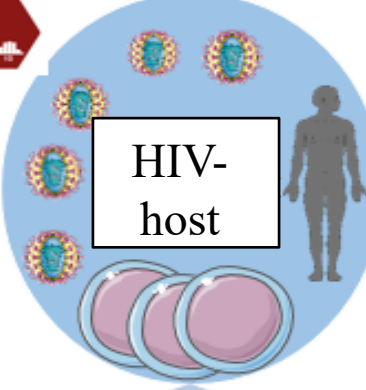
Virus-host



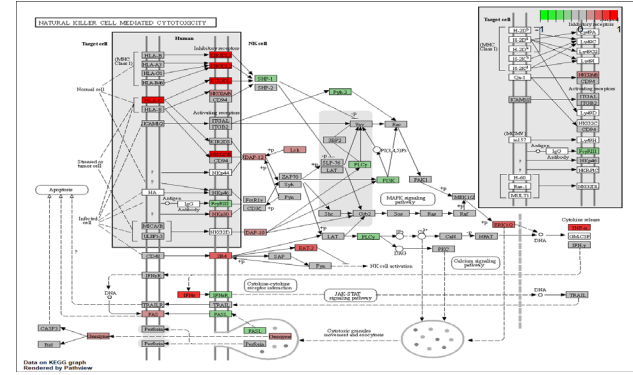
python

Stanford University
HIV DRUG RESISTANCE DATABASE

An analysis of
HIV drug
resistance and
efficacy of
ARVT



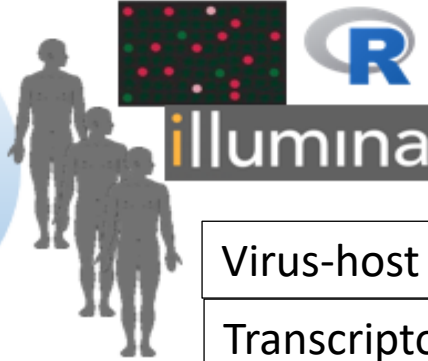
Search for master
regulators of the
long-term
nonprogression



Virus-drug
Drug-host

Genome (virus)
Transcriptome (host)

The experimental
validation of the
developed
approach



Virus-host
Transcriptome (host)

Classify nucleotide sequences of HIV protease and reverse transcriptase gene into resistant and susceptible to antiretroviral treatment using PASS approach

The classification is based on the ...

Sequence:

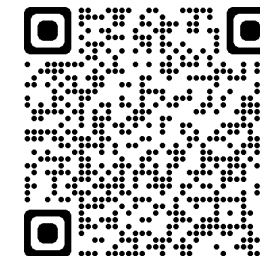
Reference sequence:

Classification:

Accession	Drug Abbreviation	Full Drug Name	ARV treatment
001	AZV	zalcitabine	zalcitabine
002	AZV	zalcitabine	zalcitabine
003	AZV	zalcitabine	zalcitabine
004	3TC	lamivudine	lamivudine
005	3TC	lamivudine	lamivudine
006	3TC	lamivudine	lamivudine
007	3TC	lamivudine	lamivudine
008	3TC	lamivudine	lamivudine
009	3TC	lamivudine	lamivudine
010	3TC	lamivudine	lamivudine
011	3TC	lamivudine	lamivudine
012	3TC	lamivudine	lamivudine
013	3TC	lamivudine	lamivudine
014	3TC	lamivudine	lamivudine
015	3TC	lamivudine	lamivudine
016	3TC	lamivudine	lamivudine
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038	3TC	lamivudine	lamivudine
039	3TC	lamivudine	lamivudine
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041	3TC	lamivudine	lamivudine
042	3TC	lamivudine	lamivudine
043	3TC	lamivudine	lamivudine
044	3TC	lamivudine	lamivudine
045	3TC	lamivudine	lamivudine
046	3TC	lamivudine	lamivudine
047	3TC	lamivudine	lamivudine
048	3TC	lamivudine	lamivudine
049	3TC	lamivudine	lamivudine
050	3TC	lamivudine	lamivudine

A principal scheme of HIV-host interaction analysis

HIV-host web resource



www.way2drug.com/hiv-host/



HIV-host

The web-platf

way2drug.com/hiv-host/nlp/

Way2Drug Home About Contact

Min: Max: CD4 cells

Column visibility Filter Excel PDF Export CD4(all) Export protein(all)

Excel PDF

Classify nucleotide sequences of HIV prot

The classification is based on the ...

Upload Sequence*

Drug's type:

Get Results Try Example

* indicates required fields

PubMed ID	Name obj 1	Age	Gender	The date of diagnosis	Therapy	Sequence	Drug Abbreviation
34201561	2-{3-[[2-{4-[4-(hydroxynitroso)phenyl]-1,3-thiazol-2-yl) hydrazin-1-ylidene] methyl]-4-methoxyphenyl} benzamide bromide	38	female	2001	3TC+LPV/r++		
		40	male	2000	EFV+d4T+dc2012); 3TC+L		
		45	male	2010	3TC+ABC+D		
		36	female	2009	3TC+LPV/r++		
		66	male	2007	3TC+ATV+ZT		
		50	male	2013	3TC+EFV+EN		
		40	female	2004	3TC+LPV/r++ 3TC+LPV/r++	seq1	FPV
		49	female	2010	ZDV+3TC+Ef DRV/r+PHAz	seq1	ATV
		54	male	2013	3TC+LPV/r++	seq1	DRV
		43	male	2003	3TC+EFV+d4 3TC+ATV+ZT EFV+LPV/r++	seq1 seq1 seq1	IDV LPV NFV
30105631	LSD1					seq1	SQV
						seq1	TPV

Showing 1 to 8 of 8 entries

Way2Drug PREDICTIVE SERVICES

Anti-HIV drug pair synergistic combinations predictor
Supported by the Russian Science Foundation (grant No. 19-75-10097)

Home Training Sets Products/Services Interpretation Contacts

Input SMILES

Input drug name

Draw Structure

View/Hide result

Lopinavir
Add structure
Examples: Acosillin

Clear data

Get HSA data

You have selected compounds: 1-2-

Pa	Pi	Delta P	Definition
0.078	0.617	-0.539	Z1
0.058	0.706	-0.648	Z2
0.715	0.071	0.644	HSA

HSA part 1 and HSA part 2 correspond to the synergistic effect based on experimentally determined values Z1 and Z2 as described in the publication by Xu Tan, and co-authors, 2012. (PubMed link)

Hiv-host includes: (1) information on extracted names of interacting HIV and human macromolecules; (2) models for predicting the efficacy of antiretroviral therapy; (3) a web service for predicting HIV drug resistance; (4) a specialised database on HIV sequences and viral load dynamics, immunograms of HIV-infected patients on specific antiretroviral therapy regimens; (5) a web resource for predicting drug synergies in inhibiting HIV replication.

<http://www.way2drug.com/hiv-host/>

Big data analysis of medical virology to find new effective and safe antiviral compounds and optimise therapy for infectious diseases

The aim of our study is to develop an *in silico* approach for the extracting knowledge about viruses and the host (the human body), and potential antiviral agents based on the mining of massive amounts of scientific publications

DrugProt, CHEMDNER

22301815 T
 22301815 T
 22301815 T
 22301815 A
 22301815 A

Annotated corpora

11 23 carbohydrate FAMILY
 65 73 cortisol TRIVIAL
 154 164 endosulfan TRIVIAL
 15 25 endosulfan TRIVIAL
 30 44 organochlorine FAMILY

Publication databases

Machine learning

Regular expressions

5-Alkyl-2-[(methylthiomethyl)thio]-6-(benzyl)-pyrimidin-4-(1H)-ones as potent non-nucleoside reverse transcriptase inhibitors of S-DABO series.

5-Alkyl-2-[(methylthiomethyl)thio]-6-(benzyl)-pyrimidin-4-(1H)-ones as potent non-nucleoside **reverse transcriptase** inhibitors of S-DABO series.

Effects of SKF 108922, an HIV-1 protease inhibitor, on retrovirus replication in mice.

Effects of **SKF 108922**, an HIV-1 **protease** inhibitor, on retrovirus replication in mice.

Pyrrolyl aryl sulfones (PASs) have been recently reported as a new class of human immunodeficiency virus type 1 (HIV-1) reverse transcriptase (RT) inhibitors acting at the non-nucleoside binding site of this enzyme.

Pyrrolyl aryl sulfones (PASs) have been recently reported as a new class of human immunodeficiency | virus type 1 (HIV-1) **reverse transcriptase (RT)** inhibitors acting at the non-nucleoside binding site of this enzyme.

Associations and relations extraction

pmid	text
16357751	Here, we report a new antiviral mechanism in which MxB restricts the nuclear import of HIV-1 regulatory protein Rev, and as a result, diminishes Rev-dependent expression of HIV-1 Gag protein. ...In addition, HIV-1 can overcome this inhibition by MxB th...
14967461	Tenofovir disoproxil fumarate (TDF) is one of the nucleotide analogs capable of inhibiting the reverse transcriptase (RT) activity of HIV and hepatitis B virus (HBV). ...These pervasive mechanisms are insufficient to prevent viral inhibition completely but may contr...
16554356	Integrase-RNA interactions underscore the critical role of integrase in HIV-1 virion morphogenesis.
16789740	Inhibition of IN-RNA interactions resulted in mislocalization of the viral ribonucleoprotein complexes

Conditional random fields,
J. Lafferty et al., 2001

Naïve Bayes,
O.Tarasova et al., 2022

BioBERT,
L. Weber et al., 2021

Naïve Bayes approach for chemical and biological NER

Class	SYSTEMATIC	Fragments of texts
Target token	cyclohexane	
Context window 1	with cyclohexane and	
Context window 2	extraction with cyclohexane and determination	
Context window 3	hydroxide extraction with cyclohexane and determination by	

“cyclohexane”

– a set of 43 multi- n -grams with $n=5$:

{A, AN, ANE, C, CL, CLO, CLOH, CLOHE, CY, CYC, CYCL, CYCLO, E, EX, EXA, EXAN, EXANE, H, HE, HEX, HEXA, HEXAN, L, LO, LOH, LOHE, LOHEX, N, NE, O, OH, OHE, OHEX, OHEXA, X, XA, XAN, XANE, Y, YC, YCL, YCLO, YCLOH}

The naïve-Bayes CNER algorithm is based on the specific B -statistics, which are calculated according to the following expressions:

$$P(C_k) = \frac{N_k}{N}, P(C_k|g_i) = \frac{N_{ik}}{N_i},$$

$$S_{0k} = 2P(C_k) - 1, S_k = \text{Sin} \left[\sum_{i=1}^m \text{ArcSin} (2(C_k|g_i) - 1) \right],$$

$$B_k = \frac{S_k - S_{0k}}{1 - S_k \cdot S_{0k}},$$

where N is the number of FoTs (tokens) in the training set and N_k is the number of FoTs belonging to the type C_k .

IAP (average), LOO CV		
	N -gram = 5	N -gram = 6
Context window = 0	0.86	0.96
Context window = 1	0.95	0.96

Verification of recognized entities. Dictionaries

Chemical named entities

- CAS common chemistry API
- ChemSpider Web API
- PubChem PUG REST
- Manually

Proteins and genes

UniProt website REST API

Diseases

Human Disease Ontology

- dictionaries can help to filter out some false positives of chemical named entities and improve accuracy of recognition;
- dictionaries can be efficiently used for recognition of diseases and disorders

Accuracy of named entity recognition

Chemicals, ML + dictionary

Validation type	Precision	Recall	F ₁ -score
5-fold CV	0.89	0.83	0.86
Manual annotation, external test	0.84	0.79	0.81

Proteins, ML + dictionary

5-fold CV	0.87	0.84	0.85
Manual annotation, external test	0.84	0.79	0.81

Diseases and disorders, ML + dictionary

5-fold CV	0.84	0.79	0.81
Manual annotation, external test	0.80	0.76	0.78

Extraction of associations between entities

1. Named entity recognition in the abstracts of relevant publications

ML-based selection of relevant publications; associations with /relations to a set of keywords characterising a set of publications belonging to a particular class

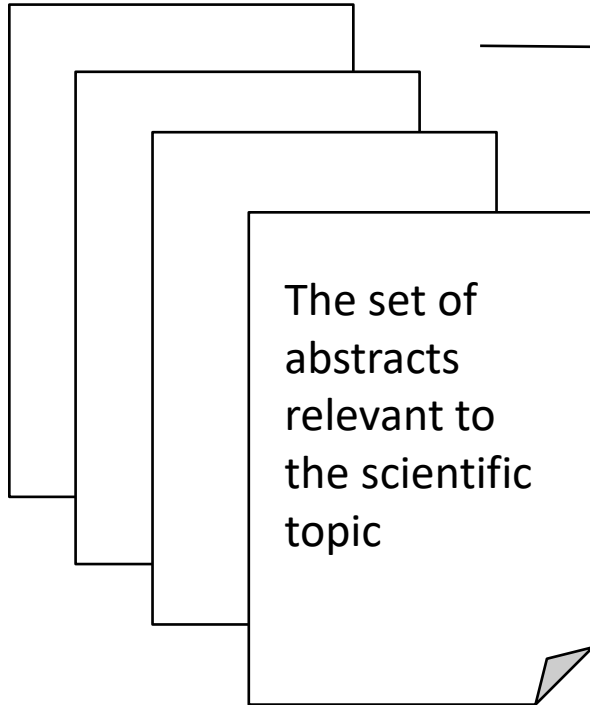
2. Rule-based approach using a set of pattern phrases

Identification of relationships in an abstract/full text or part thereof

3. Co-occurrences

Search for possible relationships that have not yet been investigated or shown in an experiment

Extraction of associations. Named entity recognition in the abstracts of relevant publications



Pyrrolyl aryl sulfones (PASs) have been recently reported as a new class of human immunodeficiency virus type 1 (HIV-1) reverse transcriptase (RT) inhibitors acting at the non-nucleoside binding site of this enzyme.

Pyrrolyl aryl sulfones (PASs) have been recently reported as a new class of human immunodeficiency virus type 1 (HIV-1) **reverse transcriptase (RT)** inhibitors acting at the non-nucleoside binding site of this enzyme.

Example: The set of proteins involved in antiviral response against HIV-1 and SARS-CoV-2

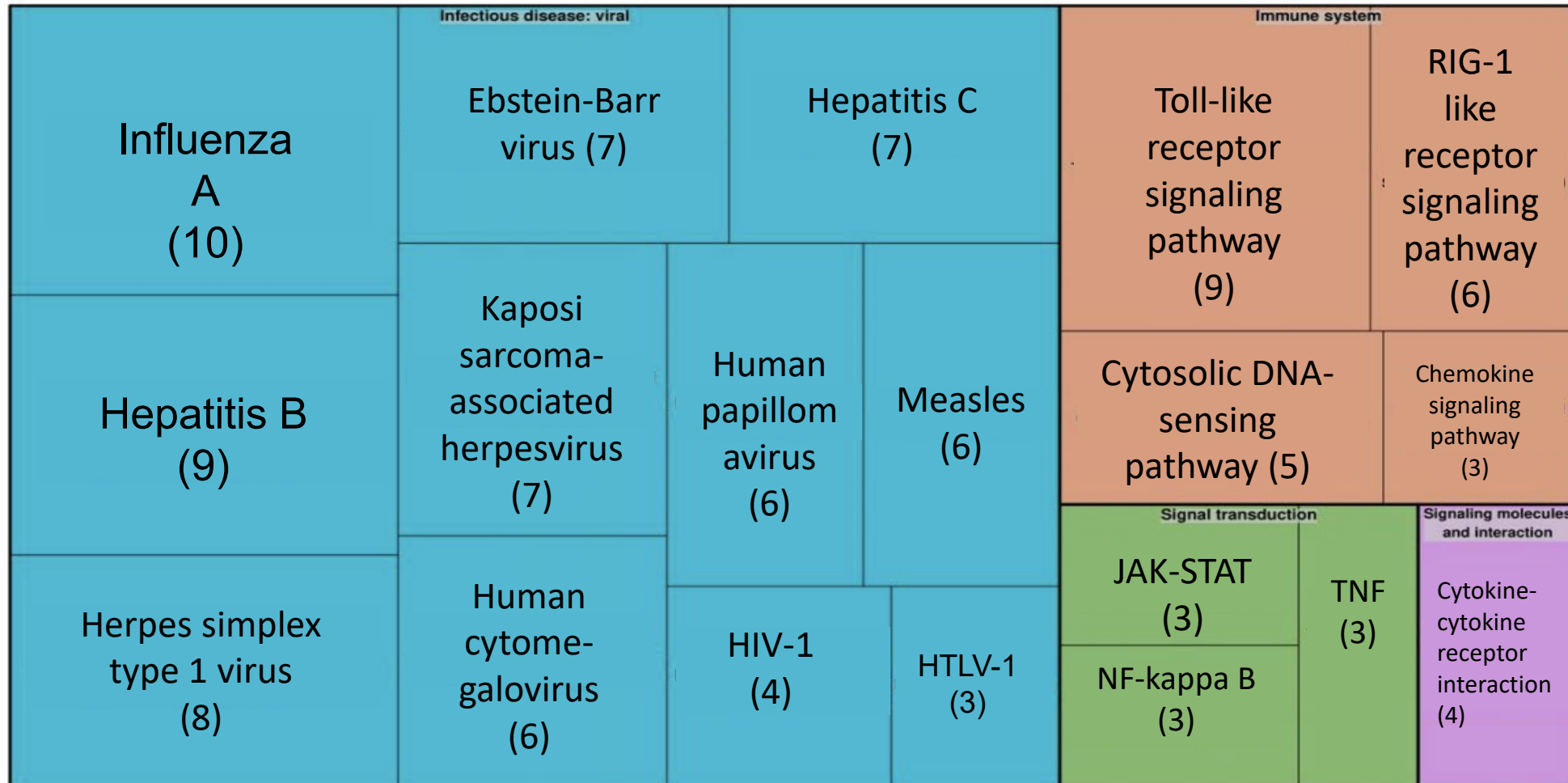
Protein Name	UniProt ID ¹	Species ²	Tissue ³	Process
AIP4	Q96J02	<i>Homo sapiens</i>	Widely expressed	Inflammation
Beclin 1	Q14457	<i>Homo sapiens</i> ¹	Ubiquitous	Autophagy of immune cells
Cathepsin B	P07858	<i>Homo sapiens</i>	Widely expressed	Entry of the virus Viral replication (HIV-1)
Cathepsin L	Q5K630	<i>Homo sapiens</i>	Widely expressed	Entry of the virus
Complement C3	P01024	<i>Homo sapiens</i>	Blood plasma and over 200 tissues	Immune response Inflammation Complement activation
IFITM1	P13164	<i>Homo sapiens</i>	Bone and over 200 tissues	Immune response

^{1,2,3} UniProt ID, species, tissue are the identifiers of proteins in UniProt database.

O.A. Tarasova et al., *Molecules*, 2020

R.M. Bonotto et al., *Antiviral Res.*, 2023; Hashimoto R. et al., *Mol Ther Nucleic Acids.*, 2021

Gene set enrichment analysis based on literature mining results



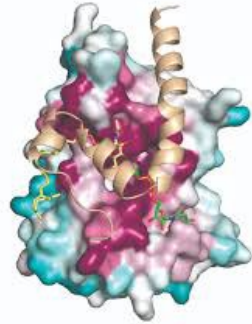
KEGG pathways enriched in the genes associated with human proteins involved in both SARS-CoV-2–host and Dengue–host interactions. Each color represents an individual pathway. The size of each box reflects the number of proteins involved in that particular pathway. The number of proteins involved in each pathway is given in brackets.

Extraction of associations. Pattern phrases



Chemical compound

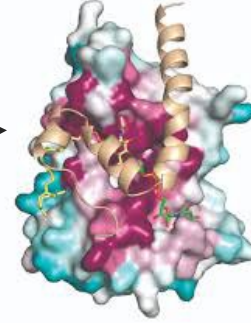
Set of pattern phrases



Protein



Protein/gene

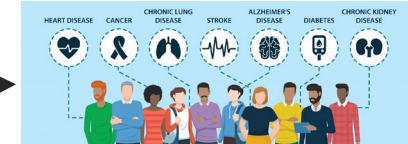


Protein/gene

Set of pattern phrases



Protein/gene



Disease

Recognition for two names (protein and disease) in one abstract selected by relevance by the particular diseases or disorders

Example: Hedgehog pathway and Cancer

Main term	Pattern	Example
interact	Interaction of P1 P2	Interaction of Vpr with other proteins such as EF2
regulate	regulation of P1 P2	Regulation of IDO by HIV-1 Tat protein
inhibit	C1 inhibits P1	RT1t49 inhibits recombinant RT

Accuracy of associations extraction: precision is 0.78, recall is 0.91, F₁-score is 0.84

Chemical named entity recognition and relation extraction

Article

Identification of Molecular Mechanisms Involved in Viral Infection Progression Based on Text Mining: Case Study for HIV Infection

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Abstract: Viruses cause various infections that may affect human lifestyle for durations ranging from several days to for many years. Although preventative and therapeutic remedies are available for many viruses, they may still have a profound impact on human life. The human immunodeficiency virus type 1 is the most common cause of HIV infection, which represents one of the most dangerous and complex diseases since it affects the immune system and causes its disruption, leading to secondary complications and negatively influencing health-related quality of life. While highly active antiretroviral therapy may decrease the viral load and the velocity of HIV infection progression, some individual peculiarities may affect viral load control or the progression of T-cell malfunction induced by HIV. Our study is aimed at the text-based identification of molecular mechanisms that may be involved in viral infection progression, using HIV as a case study. Specifically, we identified human proteins and genes which commonly occurred, overexpressed or underexpressed, in the collections of publications relevant to (i) HIV infection progression and (ii) acute and chronic stages of HIV infection. Then, we considered biological processes that are controlled by the identified protein and genes. We verified the impact of the identified molecules in the associated clinical study.

Keywords: HIV/AIDS; viral infection; viral infection progression; acute HIV infection; chronic HIV infection; text mining; machine learning



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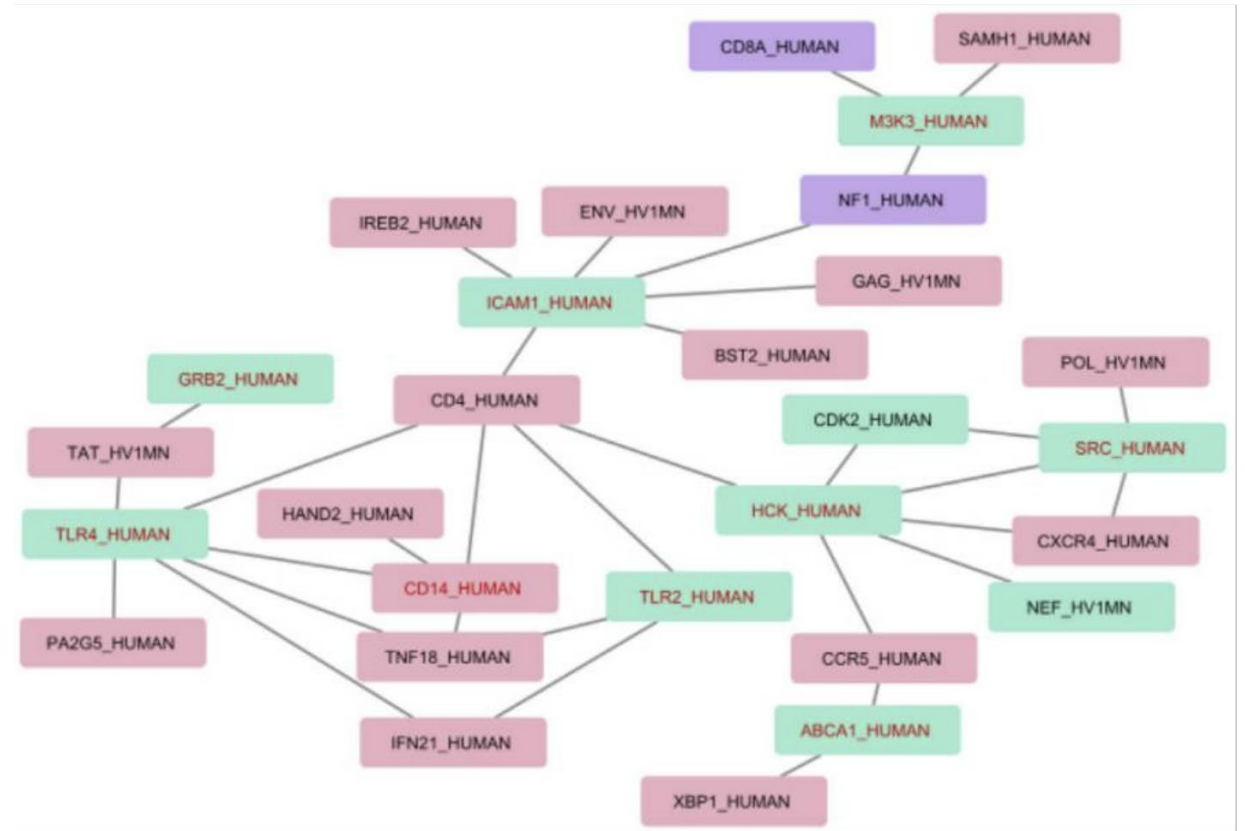
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1. Introduction

Viruses that cause human infectious diseases can cause health problems of varying severity. While some viruses, such as SARS-CoV (Severe acute respiratory syndrome coronavirus), SARS-CoV-2 (Severe acute respiratory syndrome coronavirus 2), Dengue, and Zika have gained recent global attention as they spread rapidly in their particular regions or across the world, viruses that have been intensively studied for decades, including the human immunodeficiency virus (HIV) and hepatitis C virus (HCV), still represent great challenges for humanity.

HIV-infection, the cause of acquired immunodeficiency syndrome, affects more than 38 million people globally. HIV evades and disrupts the immune system, leading to CD4+ and CD8+ T-cell depletion and chronic immune activation and inflammation [1,2]. The disruption of T-lymphocytes, inflammation, virus-induced cell death, and the apoptosis of CD4+ T-cells are the major elements of immune system dysfunction. HIV is able to escape immune response due to its high mutation rate and the ability of some mutated variants

For a variety of human genes, information about changes in expression levels during the course of HIV infection has been shown in the experiment



Red font - differentially expressed genes for which the differences are reliable and confirmed in experiment

Experimental validation of the results based on the prospective clinical study

11 patients before the start of antitertoviral therapy, (ART), peripheral blood mononuclear cells (PBMC)

9 patients after 24 weeks of ART, PBMC

Patients with HIV-infection over 1 year/
less than 1 year (5 patients/ 6 patients)

RNASeq, HiSeq1500 (Illumina)

Differences in expression level:

- 606 genes ($p < 0,1$);
- 183 genes ($p < 0.05$)

- Central Research Institute of Epidemiology, Moscow, Russia
- Krasnodar clinical center of HIV profilaxys and cure

Differences in expression level, 24 weeks after HAART: 165 genes ($p < 0.05$) compared to before treatment
16 genes ($p < 0.05$) in two groups based on the immunological effectiveness

Genes that identified in the text mining that were found to be differentially expressed in the experiment

Gene	Name	Process	PMIDs	Log ₂ FoldChange	P _{adj}
CLEC5A ↑	C-type lectin domain family 5 member A	Immune response; negative regulation of apoptotic process; negative regulation of myeloid cell apoptotic process	31867016	1.67	0.0006
TLR2 ↑	Toll-like receptor 2, CD282	Immune response Inflammatory response apoptotic process Regulation of gene expression	32093694 32516401 28730622	1.4	0.019
CD14 ↑	Monocyte differentiation antigen CD14	Immune response apoptotic process Inflammatory response	34211989 33487130	1.14	0.04
CD86 ↑	T-lymphocyte activation antigen CD86	Immune response Negative regulation of T cell proliferation	34630420	0.89	0.046
NAMPT ↑	Nicotinamide phosphoribosyltransferase	Autophagy	-	2.0	0.03

Named entity recognition and relation extraction for solving various biological tasks

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https://doi.org/10.1186/s13321-022-00633-4

Journal of Cheminformatics

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ORIGINAL RESEARCH
published: 22 December 2020
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RESEARCH

Open Access

Chemical named entity recognition in the texts of scientific publications using the naïve Bayes classifier approach

O. A. Tarasova*, A. V. Rudik, N. Yu. Biziukova, D. A. Filimonov and V. V. Poroikov



Abstract

Motivation: Application of chemical named entity recognition (CNER) algorithms allows retrieval of information from texts about chemical compound identifiers and creates associations with physical–chemical properties and biological activities. Scientific texts represent low-formalized sources of information. Most methods aimed at CNER are based on machine learning approaches, including conditional random fields and deep neural networks. In general, most machine learning approaches require either vector or sparse word representation of texts. Chemical named entities (CNEs) constitute only a small fraction of the whole text, and the datasets used for training are highly imbalanced.

Methods and results: We propose a new method for extracting CNEs from texts based on the naïve Bayes classifier combined with specially developed filters. In contrast to the earlier developed CNER methods, our approach uses the representation of the data as a set of fragments of text (FoTs) with the subsequent preparation of a set of multi-*n*-grams (sequences from one to *n* symbols) for each FoT. Our approach may provide the recognition of novel CNEs. For CHEMDNER corpus, the values of the sensitivity (recall) was 0.95, precision was 0.74, specificity was 0.88, and balanced accuracy was 0.92 based on five-fold cross validation. We applied the developed algorithm to the extracted CNEs of potential Severe acute respiratory syndrome coronavirus 2 (SARS-CoV-2) main protease (Mpro) inhibitors. A set of CNEs corresponding to the chemical substances evaluated in the biochemical assays used for the discovery of Mpro inhibitors was retrieved. Manual analysis of the appropriate texts showed that CNEs of potential SARS-CoV-2 Mpro inhibitors were successfully identified by our method.

Conclusion: The obtained results show that the proposed method can be used for filtering out words that are not related to CNEs; therefore, it can be successfully applied to the extraction of CNEs for the purposes of cheminformatics and medicinal chemistry.

Keywords: Chemical named entity recognition, CNE, CNER, Naïve Bayes classifier, SARS-CoV-2, Mpro inhibitors

Introduction

An analysis of texts is essential for extracting new knowledge about chemical compounds, drugs, targets, pathological processes and diseases; it allows determining various relationships including identification of molecular

mechanisms, pharmacological effects and toxicity of drug, pathophysiological processes and determining drug-target-disease relationships [1, 2]. Extraction of chemical named entities (CNEs) from scientific publications is an essential task since it allows using the obtained data for building chemical-target associations [3], leading to improvement of the data curation [3–6]. Chemical named entity recognition (CNER) algorithms can help create large sets of named entities of chemical compounds associated with physical and chemical properties or biological

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Automated Extraction of Information From Texts of Scientific Publications: Insights Into HIV Treatment Strategies

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Text analysis can help to identify named entities (NEs) of small molecules, proteins, and genes. Such data are very important for the analysis of molecular mechanisms of disease progression and development of new strategies for the treatment of various diseases and pathological conditions. The texts of publications represent a primary source of information, which is especially important to collect the data of the highest quality due to the immediate obtaining information, in comparison with databases. In our study, we aimed at the development and testing of an approach to the named entity recognition in the abstracts of publications. More specifically, we have developed and tested an algorithm based on the conditional random fields, which provides recognition of NEs of (i) genes and proteins and (ii) chemicals. Careful selection of abstracts strictly related to the subject of interest leads to the possibility of extracting the NEs strongly associated with the subject. To test the applicability of our approach, we have applied it for the extraction of (i) potential HIV inhibitors and (ii) a set of proteins and genes potentially responsible for viremic control in HIV-positive patients. The computational experiments performed provide the estimations of evaluating the accuracy of recognition of chemical NEs and proteins (genes). The precision of the chemical NEs recognition is over 0.91; recall is 0.86, and the F1-score (harmonic mean of precision and recall) is 0.89; the precision of recognition of proteins and genes names is over 0.86; recall is 0.83; while F1-score is above 0.85. Evaluation of the algorithm on two case studies related to HIV treatment confirms our suggestion about the possibility of extracting the NEs strongly relevant to (i) HIV inhibitors and (ii) a group of patients i.e., the group of HIV-positive individuals with an ability to maintain an undetectable HIV-1 viral load overtime in the absence of antiretroviral therapy. Analysis of the results obtained provides insights into the function of proteins that can be responsible for viremic control. Our study demonstrated the applicability of the developed approach for the extraction of useful data on HIV treatment.

Keywords: text mining, data mining, named entity recognition, NER, virus-host interactions, HIV, viremic control

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Identification of Proteins and Genes Associated with Hedgehog Signaling Pathway Involved in Neoplasm Formation Using Text-Mining Approach

Nadezhda Yu. Biziukova, Sergey M. Ivanov, and Olga A. Tarasova*

Abstract: Analysis of molecular mechanisms that lead to the development of various types of tumors is essential for biology and medicine, because it may help to find new therapeutic opportunities for cancer treatment and cure including personalized treatment approaches. One of the pathways known to be important for the development of neoplastic diseases and pathological processes is the Hedgehog signaling pathway that normally controls human embryonic development. Systematic accumulation of various types of biological data, including interactions between proteins, regulation of genes transcription, proteomics, and metabolomics experiments results, allows the application of computational analysis of these big data for identification of key molecular mechanisms of certain diseases and pathologies and promising therapeutic targets. The aim of this study is to develop a computational approach for revealing associations between human proteins and genes interacting with the Hedgehog pathway components, as well as for identifying their roles in the development of various types of tumors. We automatically collect sets of abstract texts from the NCBI PubMed bibliographic database. For recognition of the Hedgehog pathway proteins and genes and neoplastic diseases we use a dictionary-based named entity recognition approach, while for all other proteins and genes machine learning method is used. For association extraction, we develop a set of semantic rules. We complete the results of the text analysis with the gene set enrichment analysis. The identified key pathways that may influence the Hedgehog pathway and their roles in tumor development are then verified using the information in the literature.

Keywords: text-mining; data mining; Hedgehog pathway; neoplastic processes; enrichment analysis; pathology molecular mechanisms

1 Introduction

The Hedgehog (Hh) signaling pathway includes

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components that are key players in embryonic development, while it is mostly inactive in adults, excluding processes of tissue repair such as wound healing^[1,2]. However, multiple studies showed activation of proteins included in the Hh pathway in cancer development. In particular, some components of the Hh pathway may be upregulated in radio and chemo-resistant types of tumors, most of which are brain tumors^[3]. Involvement of the Hh pathway proteins in cancer progression can be explained by the convergent functions of these proteins in embryonic development and tumor formation through the regulation of proliferation, differentiation, and migration^[4].

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Extracting knowledge about viruses and the host (the human body),
and potential antiviral agents based on the mining of massive amounts
of scientific publications

Selection of antiviral compounds with direct antiviral activity



Drug indications
(Viral disease)

Drug mechanisms
(Target organism)

SARS-CoV-2, HIV, herpes simplex, hepatitis B and C, influenza

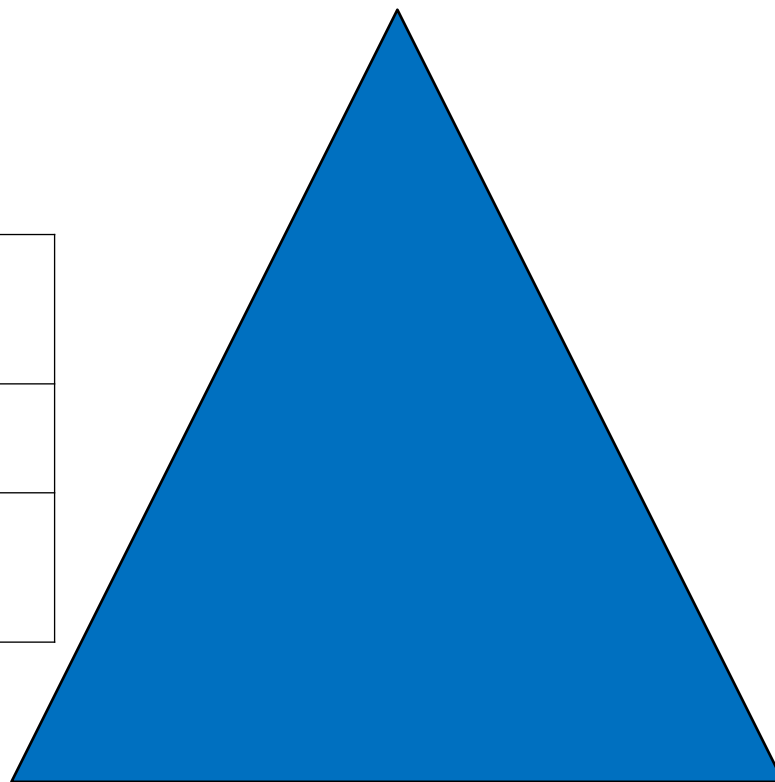
Over 170 000 abstracts

NER and relations extraction

N	Object types	Method of association extraction	Examples of relations
1	Chemical-Chemical	Pattern phrases	Combinations of drugs used in therapy; drug effects on biochemical parameters; drug-drug interactions; metabolism and biotransformation
2	Chemical-Gene	Pattern phrases	Effect of drug on biochemical parameters; effect on protein/gene
3	Chemical-Disease	Pattern phrases	Side effects and toxicity; use in therapy; use in prevention; possible biomarkers of disease
4	Chemical-miRNA	Co-occurrence	Effects of chemical compounds on molecular mechanisms
5	Chemical-SNP	Pattern phrases; Co-occurrence	Relationship between amino acid/nucleotide substitution and drug resistance/susceptibility
6	Chemical-Genotype	Pattern phrases	(Typically associations of HLA genotypes with drug hypersensitivity)
7	Gene-Gene	Pattern phrases	Interactions between proteins (including signaling pathways); part-to-part associations (protein-family)
8	Gene-Disease	Pattern phrases	Possible biomarkers; molecular mechanisms of pathogenesis; use in therapy
9	Gene-miRNA	Co-occurrence	Participation in regulation
10	Gene-SNP	Co-occurrence	Which protein substitutions lead to resistance/susceptibility to the drug
11	Disease-Disease	Pattern phrases	Main disease-symptom; comorbidities; multicomponent diseases
12	Disease-miRNA	Co-occurrence	Involvement in pathogenesis
13	Disease-SNP	Co-occurrence	Associations of changes in proteins/genes with disease manifestation and pathogenesis
14	Disease-Genotype	Pattern phrases; Co-occurrence	Other reactions and pathological processes associated with HLA genotypes

NER and extracting associations or relations

Virus



Chemicals

Host (homo sapiens)

Object types	Number of recognized (unique)	Found in databases (unique)
Gene	15 844	4 034
miRNA	640	-
Disease	55 080	7 998
SNP	10953	-
Associations / relations	Gene-Disease; Gene-gene; Gene-miRNA; Chemical-Genotype	

Object types	Number of recognized (unique)	Found in databases (unique)
Chemical named entities	83 571	6 972
Associations / relations	Chemical-miRNA; Chemical-Genotype; Chemical-chemical	

Associations / relations	Chemical-Disease; Chemical-Gene; Chemical-miRNA
--------------------------	---

Virus-host-chemicals interactions based on textome analysis for several viruses.

Preliminary results

Way2Drug Home About Contacts

Name obj 1	Type obj 1	Name obj 2	Type obj 2	Interaction
(R)-VALGANCICLOVIR	Chemical	CMV disease	Disease	Used in therapy of

Name Obj1 : (R)-VALGANCICLOVIR
TextNames Obj1 : valganciclovir
Name Obj2 : CMV disease
TextNames Obj2 : CMV disease
PMedIDs : Total documents : 38

Click to view 3 random documents Click to view all documents

8339135; 24334950; 19790144;
 33664179 21175987 15367218 15628830 17175363 17297403 17889146 18044717
 8339135 18590496 18674820 19436751 19642123 20053269 20230540 21239188
 22139888 24182826 24334950 29215460 30241930 30556615 33744428 35107143
 37279999 18171208 18236404 18279073 19422345 19704173 24983306 25560711
 25738669 30643845 36396949 19790144 26900992 17157554

Cytomegalovirus reactivation following hematopoietic stem cell transplantation. There is a high prevalence of cytomegalovirus (CMV) seropositivity in developing countries. An apparent risk of CMV reactivation increases following hematopoeitic stem cell transplantation. With effective surveillance and timely treatment using anti-viral therapy, morbidity and mortality associated with CMV reactivation can be reduced. To evaluate the incidence and morbidity associated with CMV reactivation following hematopoeitic stem cell transplantation. We retrospectively analysed 136 hematopoeitic stem cell transplant recipients at our centre for CMV reactivation and their complications. Quantification of CMV-DNA was done by PCR. CMV disease was confirmed histologically via CMV inclusion bodies or immunostaining of biopsy of the affected organ, mainly the gastrointestinal tract. A total of 13 out of 136 patients (9.56%) had CMV reactivation. 6 out of 13 patients had CMV disease, 3 of which died (23.1% of patients with CMV reactivation). CMV reactivation occurred at a median duration of 52.5 days post transplantation (range 35-178 days). The gastrointestinal tract was the organ most commonly affected by CMV. The median follow-up was 14 months (range 6 - 64 months). Through a higher rate of sero-prevalance in developing countries, the incidence of CMV infection following hematopoeitic stem cell transplantation is comparable to that reported in Western literature. Oral valganciclovir was an effective pre-emptive therapy for CMV disease.

(R)-VALGANCICLOVIR	Chemical	CMV disease	Disease	Increases risk of
--------------------	----------	-------------	---------	-------------------

SARS-CoV-2, HIV-1, hepatitis C and B, influenza, Zika, Dengue, Western Nile

Examples of relations between identified objects extracted using literature mining

Way2Drug Home About Contacts

Excel PDF

Name obj 1	Type obj 1	Name obj 2	Type obj 2	Status
DEXTROSE	Chemical	Insulin	Gene	verified
ZIDOVUDINE	Chemical	VIRAL REVERSE TRANSCRIPTASE	Gene	verified
ETOPOSIDE	Chemical	topoisomerase II	Gene	verified
DEXTROSE	Chemical	Insulin	Gene	unverified
DOLUTEGRAVIR	Chemical	VIRAL INTEGRASE	Gene	unverified
ALANINE	Chemical	ALT	Gene	unverified
GLUTATHIONE	Chemical	Catalase	Gene	unverified
ALANINE	Chemical	aspartate aminotransferase	Gene	unverified
STAVUDINE	Chemical	VIRAL REVERSE TRANSCRIPTASE	Gene	unverified
RIBAVIRIN	Chemical	Peg-IFN	Gene	unverified

Showing 1 to 10 of 560 entries (filtered from 5,683 total entries)

Previous 1 2 3 4 5 ... 56 Next

- Are used in combination
- Has positive effect on
- Has unspecified effect on
- Increases risk of
- Interacts
- Interacts with
- Investigated regarding
- Is a marker of
- Is a prodrug of
- Lead to reduction of
- Leads to
- Leads to a decrease of
- Leads to an increase of
- Leads to an unspecified effect
- May cause
- May lead to resistance to
- NS
- Negative regulation of
- Positive effect

<https://www.way2drug.com/viruses/nlp/>

Examples of amino acid substitutions associated with viral drug resistance

Way2Drug Home About Contacts

Excel PDF Search:

Name obj 1	Type obj 1	Name obj 2	Type obj 2	Relation	Status
OSELTAMIVIR CARBOXYLATE	Chemical	H275Y	SNP	NS	unverified
LAMIVUDINE	Chemical	M184V	SNP	NS	unverified
OSELTAMIVIR CARBOXYLATE	Chemical	H275Y	SNP	Resistant due to	unverified
OSELTAMIVIR CARBOXYLATE	Chemical	H274Y	SNP	NS	unverified
LAMIVUDINE	Chemical	M184V	SNP	Resistant due to	unverified
TENOFOVIR	Chemical	K65R	SNP	NS	unverified
NEVIRAPINE	Chemical	K103N	SNP	NS	unverified
OSELTAMIVIR CARBOXYLATE	Chemical	H274Y	SNP	Resistant due to	unverified
RIBAVIRIN	Chemical	rs12979860	SNP	NS	unverified
EMTRICITABINE	Chemical	M184V	SNP	NS	unverified

Name obj 1 Name obj 2 SNP

Showing 1 to 10 of 397 entries (filtered from 5,683 total entries)

Previous 1 2 3 4 5 ... 40 Next

Conclusions

- We developed the approach to extract associations between automatically recognised entities corresponding to (a) chemical named entities; (b) proteins, genes, miRNAs; (c) diseases and disorders.
- The developed approach and algorithms were validated in several studies, including the identification of genes associated with HIV infection progression and therapeutic outcome; the search for proteins and genes involved in neoplasm development and associated with the Hedgehog pathway.
- Based on the developed approach, we created an automated pipeline aimed at extracting knowledge about viruses and the host (the human body) and potential antiviral agents based on the mining of massive amounts of scientific publications.

Selected publications

- **Tarasova, O.**, Biziukova, N., Shemshura, A., Filimonov, D., Kireev, D., Pokrovskaya, A., Poroikov, V. Identification of Molecular Mechanisms Involved in Viral Infection Progression Based on Text Mining: Case Study for HIV Infection. *International Journal of Molecular Sciences.*; 2023. 24(2), 1465; <https://doi.org/10.3390/ijms24021465>
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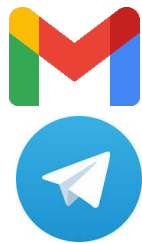
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