

Drug Repurposing: New Uses for Old Drugs or Systems Biomedicine?

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THE 5TH ANNUAL
**Drug Repositioning,
Repurposing
and Rescue Conference**



Chicago, Illinois USA
June 21-22, 2016

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Featured Speakers

Matthew DeSilva, CEO, Notable Labs

Daphna Laifenfeld, Ph.D., Director, Personalized and Predictive Medicine and Big Data Analytics, **Teva Pharmaceuticals**

Aytekin Oto, MD, Professor of Radiology, Section Chief, Abdominal Imaging, Chief of Body MRI, **The University of Chicago**

Larry Sklar, Ph.D., The Maralyn S. Budke and Robert E. Anderson Distinguished Endowed Chair in Cancer Drug Discovery, Director, UNM Center for Molecular Discovery, **The University of New Mexico School of Medicine**

Marty St. Clair, Ph.D., Clinical Virology, **ViiV Healthcare**

About the Conference

Join us in Chicago, where we will highlight the latest developments in the fields of drug repositioning, repurposing and rescue. This conference continues to serve as a global meeting place for those engaged in efforts to further drug development through new means of collaborations, including patient advocacy efforts and industry/academic/government cooperation.

Key Themes at This Year's Conference

PATIENT ADVOCACY EFFORTS

Emphasis on and engagement with patient advocacy groups, who are investing in drug repositioning efforts to an unprecedented degree

NEW PARTNERSHIPS

The conference will explore how new partnerships between various groups, including government, industry and academia are teaming up to advance repurposing efforts

COMMERCIAL CASE STUDIES

Leaders in drug repositioning will discuss their successes, failures and the way forward

COLLABORATIVE EFFORTS

Government/Academic/Industry Collaborations will be explored and highlighted in order to determine how

<http://www.drugrepositioningconference.com/index/>

Drug Repurposing, Rescue, and Repositioning



Henry Aasen Liebert, Inc. is publisher
www.liebertpub.com

A New Journal for the Drug Repurposing Community

Hermann A.M. Mucke, PhD

European Editor, Drug Repurposing, Rescue, and Repositioning.
H.M. Pharma Consultancy, Wien, Austria.

Dear reader:

What you are holding in your hand—or what you are looking at on your screen—is the premier issue of the first journal that is exclusively dedicated to new medical uses of known pharmaceutically active compounds: *Drug Repurposing, Rescue, and Repositioning*.

So, another peer-reviewed journal for the medical sciences. Why should this be necessary? Hundreds exist already.

INTERDISCIPLINARY BROADNESS DEMANDS HIGH-LEVEL INTEGRATION

To be sure, it is not as if there were no proper opportunities to publish quality articles addressing drug repurposing. Pertinent articles appear in life sciences journals that specialize in medicinal chemistry, systems biology, molecular modeling,

has been missing until now. The product you are looking at is the first coordinated and well-supported attempt to remedy this.

OPTIMAL RESOURCE UTILIZATION IS NOT RECYCLING

Several common myths need to be dispelled before experts from so many diverse fields can collaborate with maximum efficacy. Number one is that drug repurposing, rescue, and repositioning is an inherently defensive concept, promoted by pharmaceutical companies to recoup at least part of their investments in the development of their failed late-stage drug candidates, or in drugs that had to be removed from the market for safety reasons. While such things do happen, this is only the “rescue” part of the story—and probably the least significant one in economic terms.

Nor is the *repositioning* of marketed drugs something as simple as what business developers call a line extension—such as expanding the approval of a cancer drug to include additional tumor types. Rather, drug repositioning implies the use in a different disease class, and while this often exploits

Drug Repurposing, Rescue, and Repositioning

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Current Volume: 1

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Astellas continues IT-enabled Drug Repurposing Deal Drive with Excelra hook-up

June 10th 2016, Posted By: [Drug Repurposing Portal](#)



Astellas Pharma has struck its third drug repurposing agreement of the past 6 months. The latest collaboration sees Astellas start working with Excelra, an Indian informatics company that has landed 8 similar deals on the strength of its drug repurposing database and accompanying algorithms. For Excelra, the deal with Astellas marks an advance in its attempts to establish itself as a standalone business.

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Why DRP? - Society point of view.

Several benefits could arise from repurposing of the launched drugs, such as:

- finding new therapies for unmet medical needs;
- finding more efficacious therapies;
- replacing expensive with inexpensive drugs;
- substituting safer drugs for drugs with unwanted effects.

National Comprehensive Cancer Network (NCCN) estimated that **50-75% of drugs have been used through off-label prescription in USA** (Drug Discovery Today, 2014, 19: 637-644).

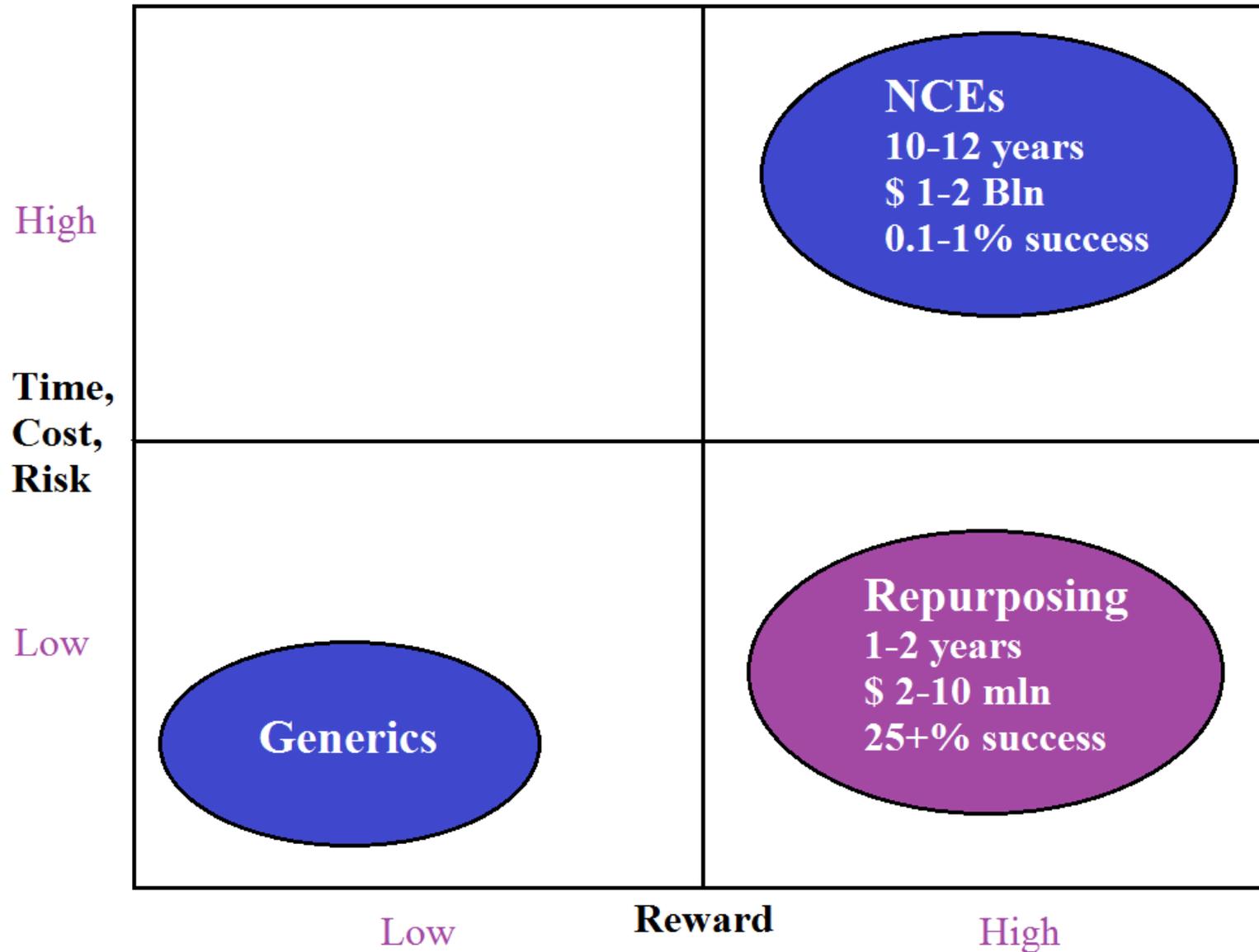
National Center for Advancing Translational Sciences (NCATS, NIH) has invested \$575 million budget on drug rescuing and repurposing.

Center for World Health & Medicine (CWHM, NIH) has initiated to provide a screening platform for identification of drugs for rare/neglected diseases (Sci. Translat. Med., 2011, 3: 80ps16).

Why DRP? - Industry point of view.

- **Successfully repositioned drugs enter the market 3-5 years faster than a conventionally developed drug and as a consequence generate income sooner.**
- **Success rates for repurposed drugs are higher and costs are lower than *de novo* R&D.**
- **It is estimated that over 2,000 failed drugs are sitting on companies shelves and that this number grows at the rate of 150-200 drugs per year.**
- **The science to evaluate new diseases continues to evolve so that science led repurposing (rather than random screening) is a viable business model.**
- **Repositioning is expected to generate up to \$20 billion in annual sales in 2012.**

DRP: Time/Cost/Risk values

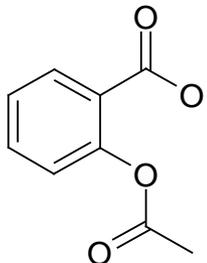


Some examples of drug repurposing

Drug

Primary indication (year)

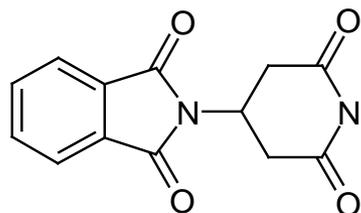
Repurposed indication (year)



Acetylsalicylic acid

NSAID (1897)

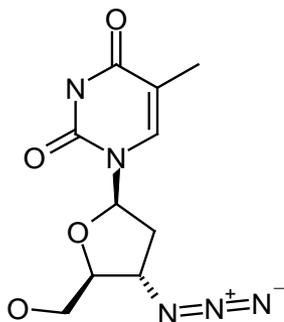
Antithrombotic (1956)



Thalidomide

Sedative (1957)

Antileprosy (1998), Antitumor (2006)

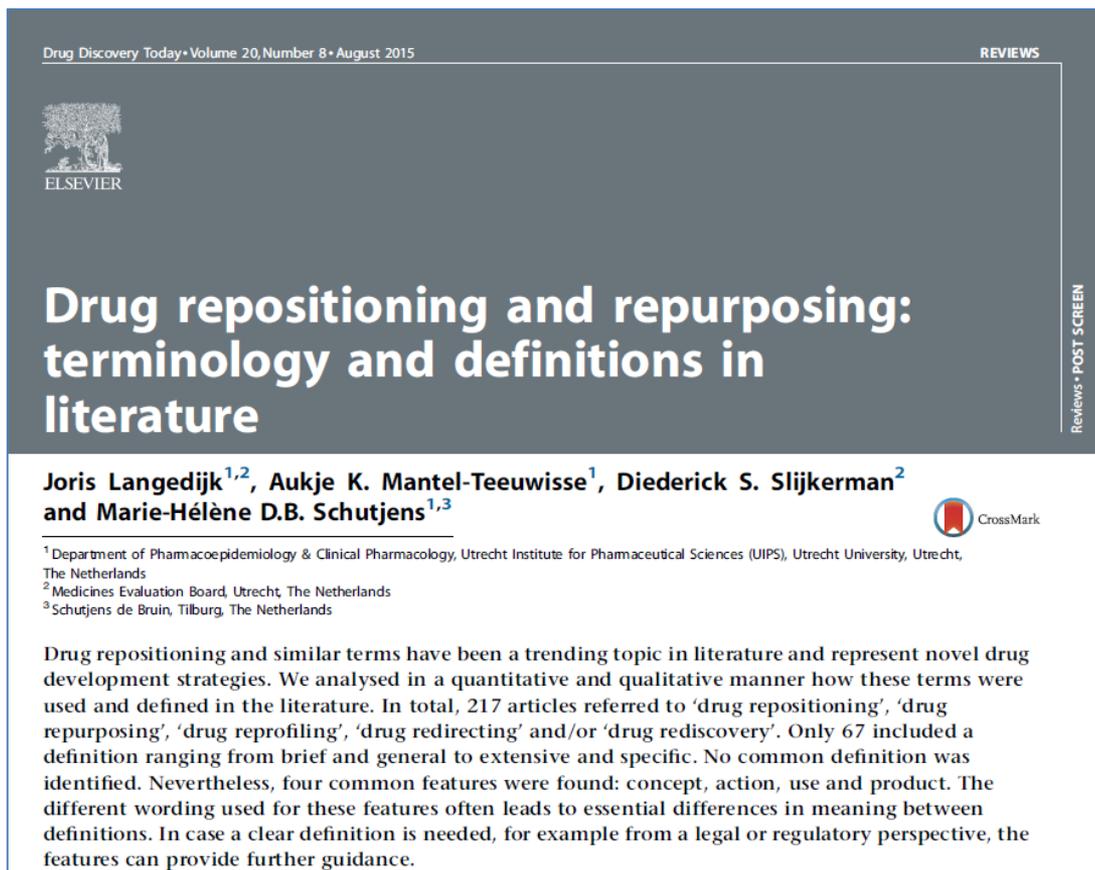


Zidovudine

Cancer (1964)

HIV/AIDS (1987)

Drug repurposing (DRP): terminology and definitions



Drug repositioning

OR

Drug repurposing

OR

Drug reprofiling

OR

Drug redirecting

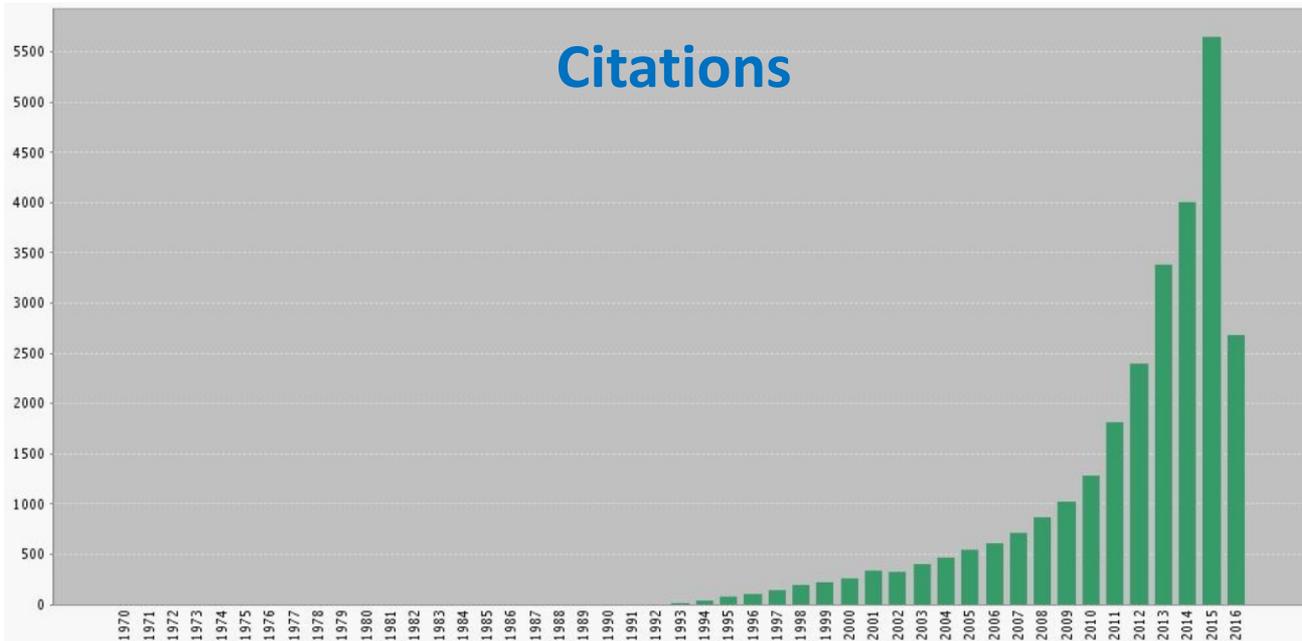
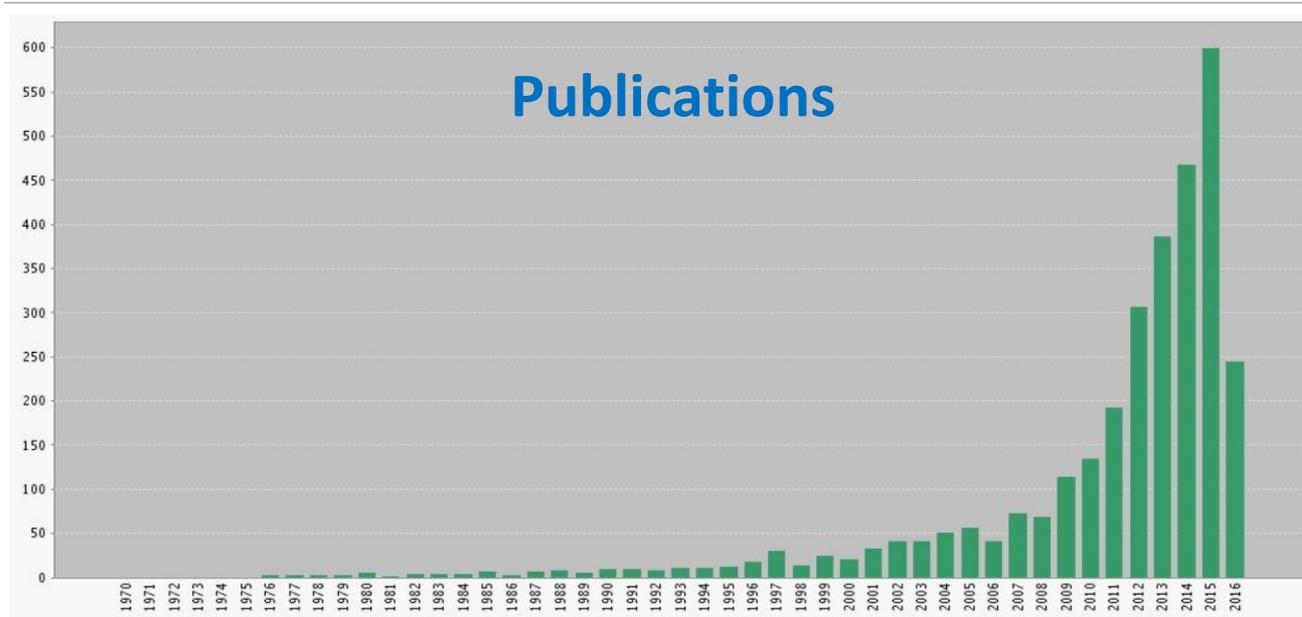
OR

Drug rediscovery

No common definition was identified. Nevertheless, four common features were found: concept, action, use and product. The different wording used for these features often leads to essential differences in meaning between definitions.

Before 2004 no articles about drug repositioning were found started to increase after 2010 in particular.

Search for DRP in Web of Science core collection



Some examples of the relevant publications

Geriatrics. 1969 Jul;24(7):113-6.

Phentolamine--rediscovery of an old drug.

Gould L.

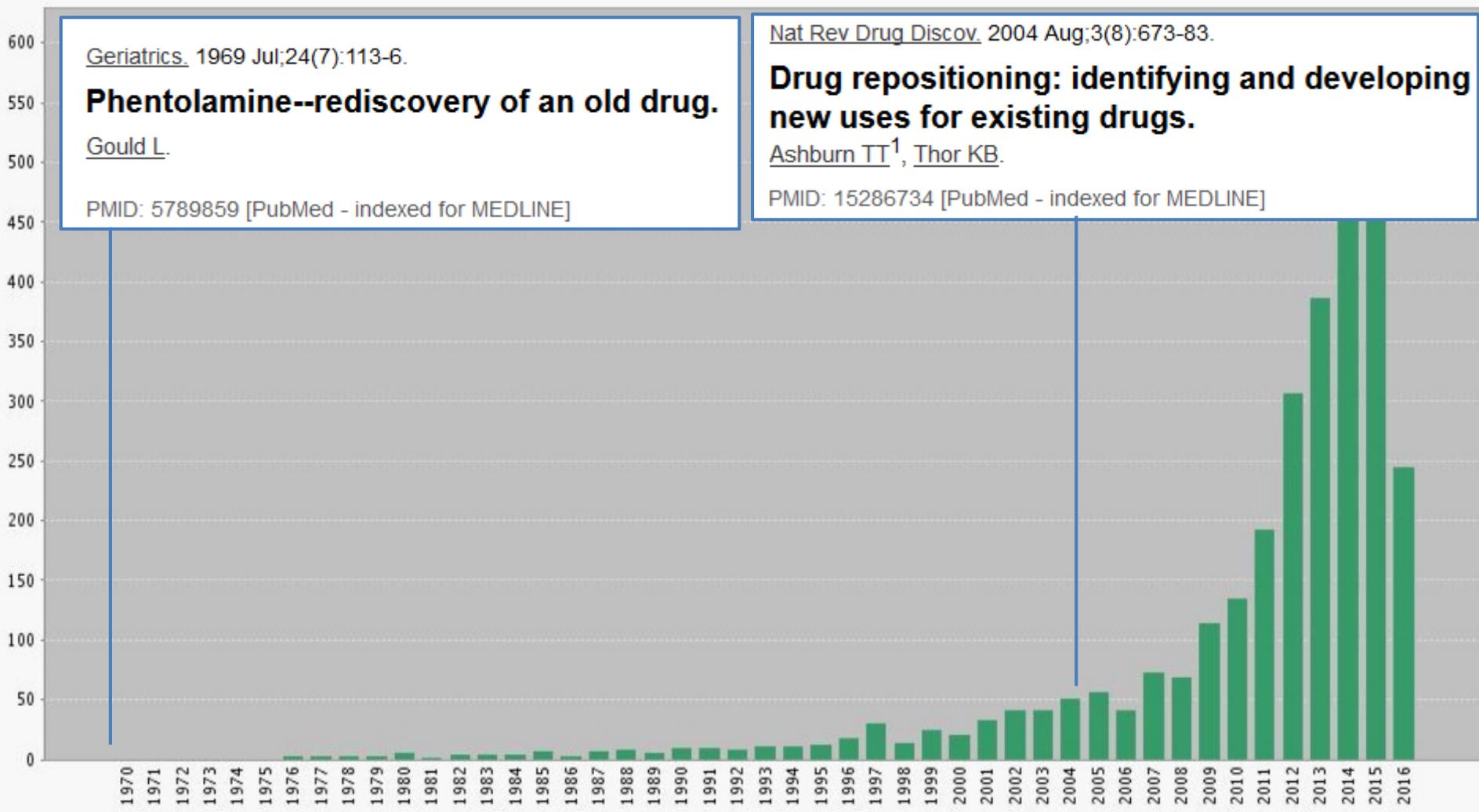
PMID: 5789859 [PubMed - indexed for MEDLINE]

Nat Rev Drug Discov. 2004 Aug;3(8):673-83.

Drug repositioning: identifying and developing new uses for existing drugs.

Ashburn TT¹, Thor KB.

PMID: 15286734 [PubMed - indexed for MEDLINE]



Publications on drug repurposing covered by Web of Science

Phentolamine— rediscovery of an old drug

LAWRENCE GOULD, M.D.

Director, Cardiac Catheterization Laboratory

Misericordia-Fordham Hospital

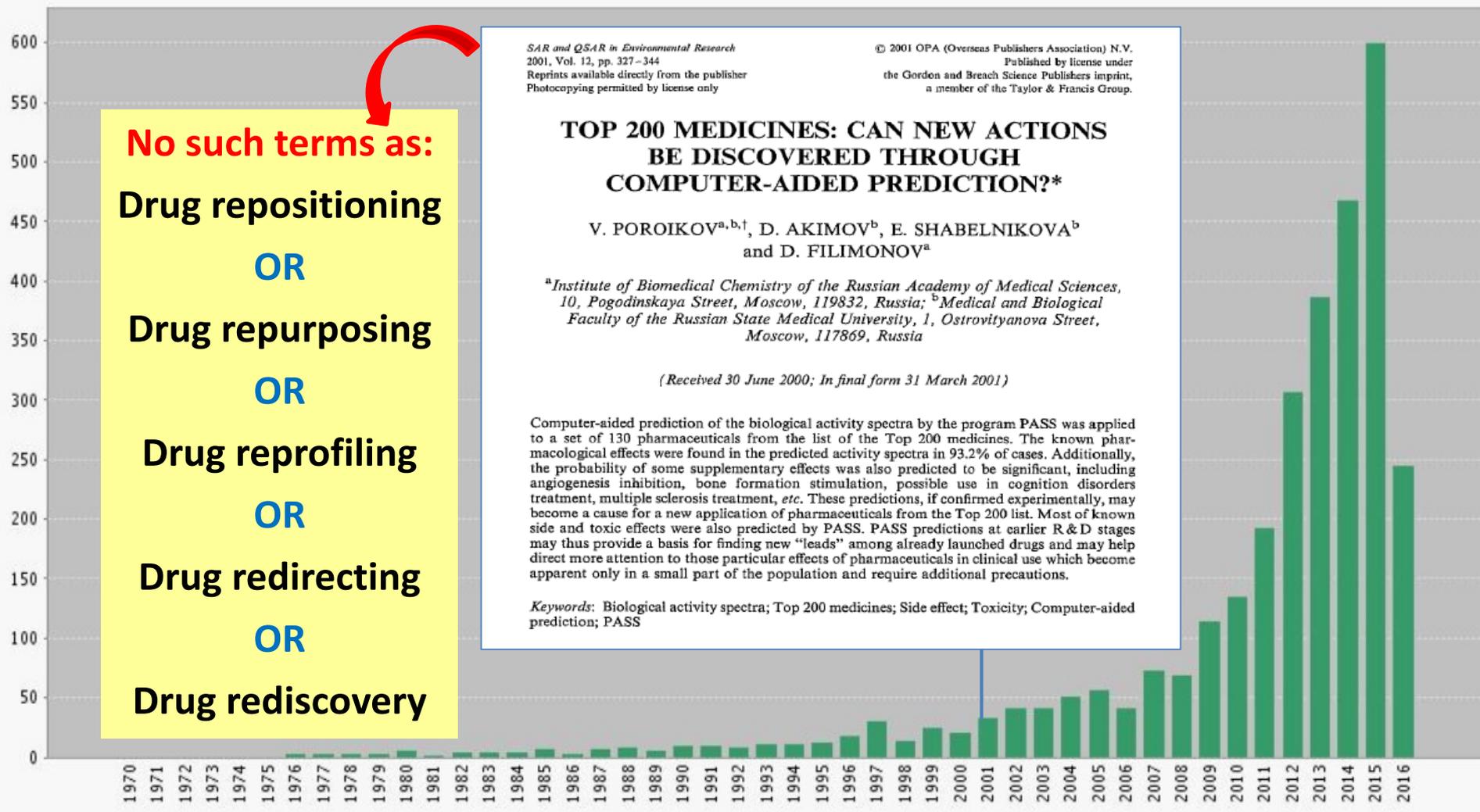
BRONX, NEW YORK

Phentolamine has long been considered to be an alpha-adrenergic blocking agent which produces arteriolar dilation unaccompanied by any primary cardiac effect. It is primarily used as a screening test for the detection of pheochromocytoma. Recent work in our laboratory has demonstrated that the drug has far greater clinical application.

GERIATRICS, *July 1969* 113

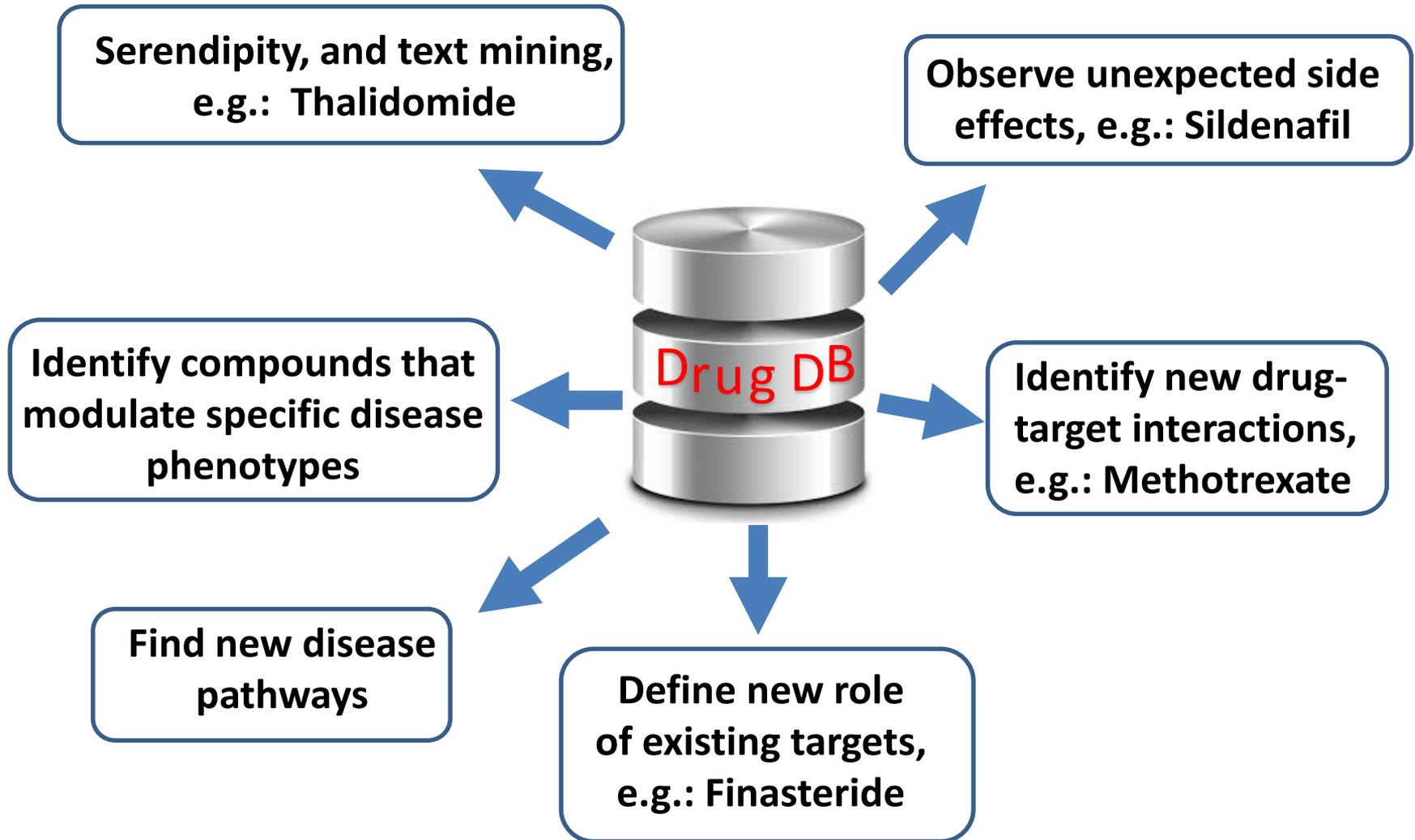
Thanks to the favor of Marc Nicklaus, CADD Group, LCB, CCR, NCI/NIH.

More examples of the relevant publications



Publications on drug repurposing covered by Web of Science

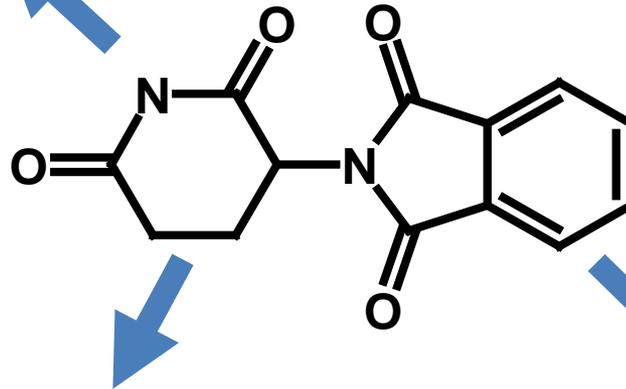
DRP: How it happens?



Thalidomide: discoveries by serendipity

Sedative, Morning sickness in pregnant women treatment - 1957

Erythema nodosum laprosum
Treatment (agonizing inflammatory condition of leprosy) – 1998 (1964)



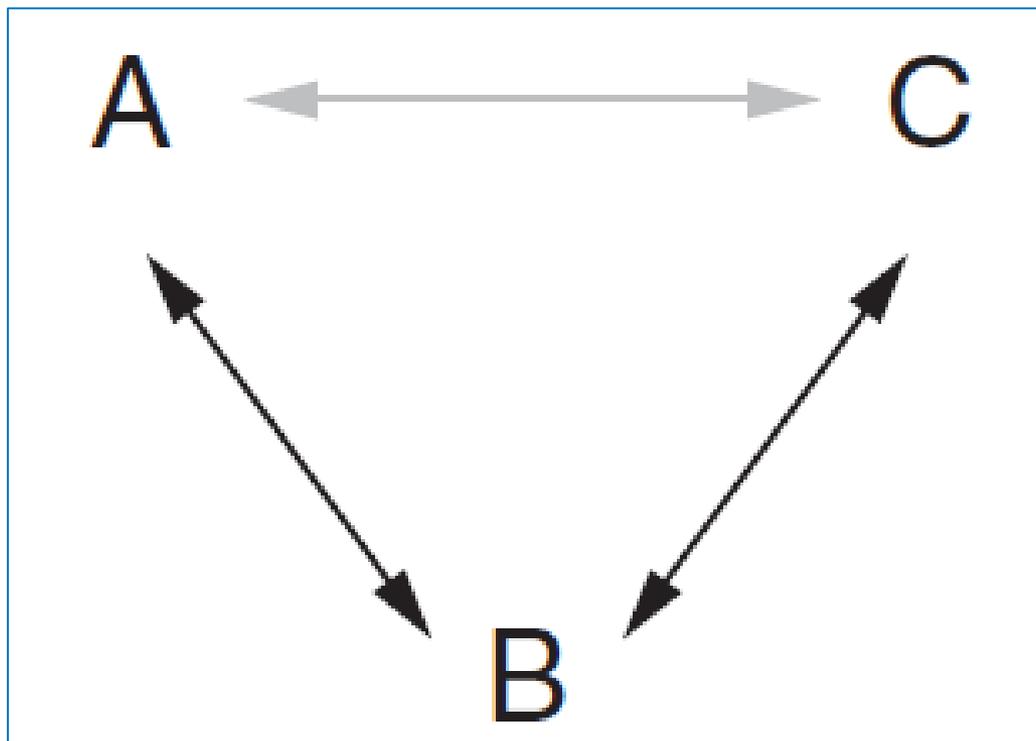
Teratogenic, Skeletal birth defects in children – early 60s (Withdrawn)

Antiangiogenic – 1994; Multiple myeloma off-the-label treatment - 1998

Baek M.-C. et al. Pharmacol. Res., 2015, 99: 185–193.

Ashburn T.T., Thor K.B. Nat. Rev. Drug. Discov, 2004, 3: 673-683.

Text mining: Literature-based discovery



Swanson's ABC model of discovery.

If concepts A and B are reported to be related to one set of publications and concepts B and C are reported to be related to another set, then A and C might be indirectly related to each other.

Swanson, D.R. (1990) Medical literature as a potential source of new knowledge. *Bull. Med. Libr. Assoc.* 78, 29–37.

Case Report ■

Generating Hypotheses by Discovering Implicit Associations in the Literature: A Case Report of a Search for New Potential Therapeutic Uses for Thalidomide

MARC WEEBER, PHD, REIN VOS, MD, PHD, HENNY KLEIN, PHD,
LOLKJE T. W. DE JONG-VAN DEN BERG, PHD, ALAN R. ARONSON, PHD,
GRIETJE MOLEMA, PHD

We find solid bibliographic evidence suggesting that thalidomide might be useful for treating acute pancreatitis, chronic hepatitis C, *Helicobacter pylori*-induced gastritis, and myasthenia gravis. However, experimental and clinical evaluation is needed to validate these hypotheses and to assess the trade-off between therapeutic benefits and toxicities.

Search in PubMed provides some evidences

NCBI Resources How To Sign in to NCBI

PubMed.gov
US National Library of Medicine
National Institutes of Health

PubMed Search

Advanced Help

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Review
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Selected items

Items: 3

- [Thalidomide alleviates acute pancreatitis-associated lung injury via down-regulation of NFκB induced TNF-α.](#)
Lv P, Li HY, Ji SS, Li W, Fan LJ.
Pathol Res Pract. 2014 Sep;210(9):558-64. doi: 10.1016/j.prp.2014.04.022. Epub 2014 May 16.
PMID: 24939146
[Similar articles](#)
- [Protective Effect of Thalidomide on Liver Injury in Rats with Acute Pancreatitis via Inhibition of Oxidative Stress.](#)
Lv P, Fan LJ, Li HY, Meng QS, Liu J.
Ann Clin Lab Sci. 2015 Fall;45(5):508-14.
PMID: 26586701
[Similar articles](#)
- [The effect of thalidomide on experimental autoimmune myasthenia gravis.](#)
Crain E, McIntosh KR, Gordon G, Pestronk A, Drachman DB.
J Autoimmun. 1989 Apr;2(2):197-202.
PMID: 2788425
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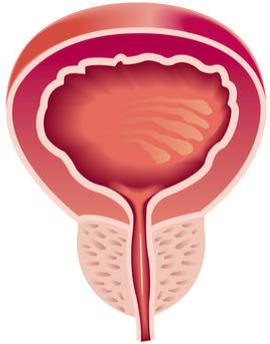
[Turn Off](#) [Clear](#)

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- [thalidomide AND \(acute pancreatitis OR chronic hep PubMed](#)
- [Generating Hypotheses by Discovering Implicit Associations](#)
- [Effect of acetazolamide on the anticonvulsant potency of se PubMed](#)

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Define new role of existing targets: Finasteride

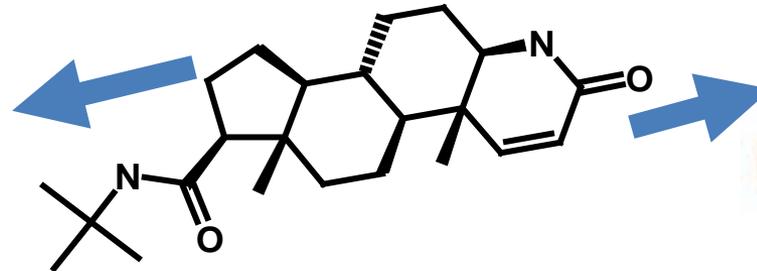
5-alpha-reductase inhibitor, Benign prostatic hyperplasia - 1992 (Proscar; Merck)



Normal prostate



Prostatic hypertrophy

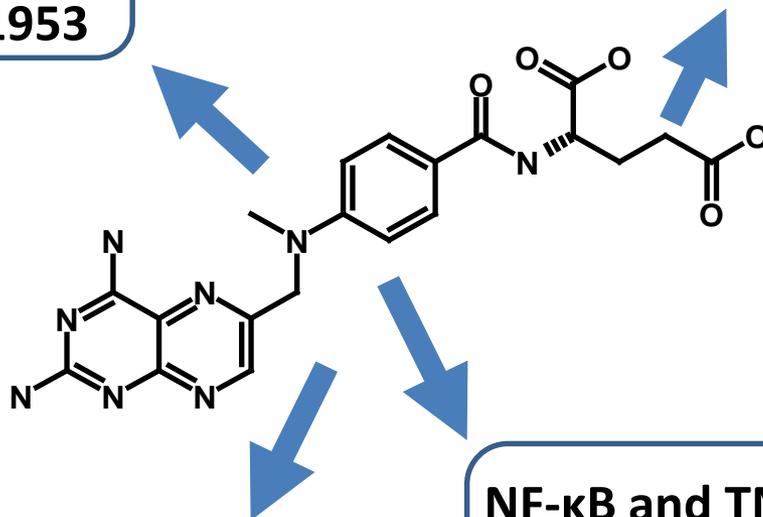


5-alpha-reductase inhibitor, Hair loss treatment - 1997
Propecia (with a fivefold lower dose), had worldwide sales
of US \$239 million in 2003

One of the major reason for DRP is drug promiscuity: Methotrexate as an example

Antineoplastic (Acute leukemia), Dihydrofolate reductase inhibitor - 1953

Antiarthritic - 1954



Osteosarcoma, breast cancer, acute lymphoblastic leukemia, and Hodgkin lymphoma - 1988

NF- κ B and TNF- α signaling pathway inhibitor, antiangiogenic, antiinflammatory - 2010

Multitargeted Drugs: The End of The “One-Target-One Disease Philosophy?”

update | discussion forum

DDT Vol. 9, No. 19 October 2004

For many years, clinicians have treated patients by combinations of drugs with different pharmacotherapeutic actions. It is being recognized that a balanced modulation of several targets can provide a superior therapeutic effect and a favourable side effect profile compared to the action of a selective ligand.

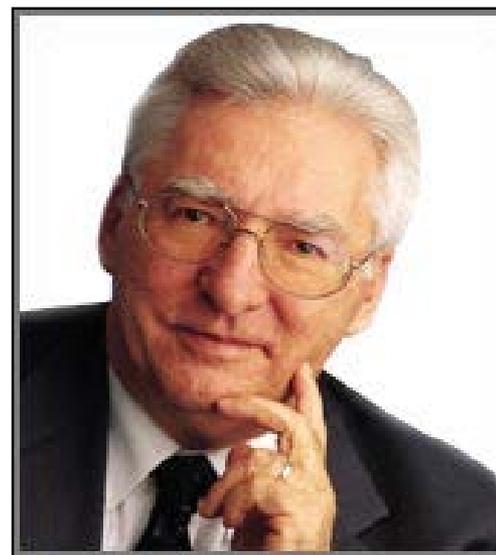
In a recent issue of *Drug Discovery Today*, Morphy *et al.* [1] discuss the opportunities and advantages associated with the design of ligands that act on two (or more) specific targets in an article entitled ‘From magic bullets to designed multiple ligands’.

Several highly specific drugs that have only one target have clearly proven the usefulness of monotarget medicine.

conjunction with amoxicillin, and in the treatment of Parkinson’s disease, where L-4-dihydroxyphenylalanine (DOPA) is concomitantly administered with DOPA-decarboxylase and catechol-O-methyltransferase inhibitors. The risk with combination therapies is that the use of multiple drugs introduces problems with pharmacokinetics, toxicity and patient compliance. To circumvent these difficulties, and after

a truly rational computer-generation of DM ligands should not be based on the interactions of structural elements, but rather the comparison and association of true pharmacophores.

The second approach (screening approach) to DM ligands is based on the screening of large libraries for the two relevant bioassays. The substantial screening of a large number of compounds, which therefore have a

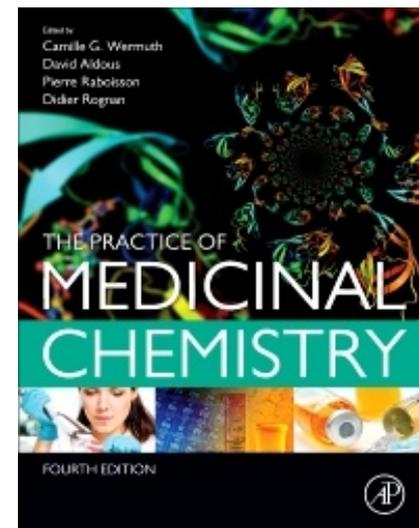
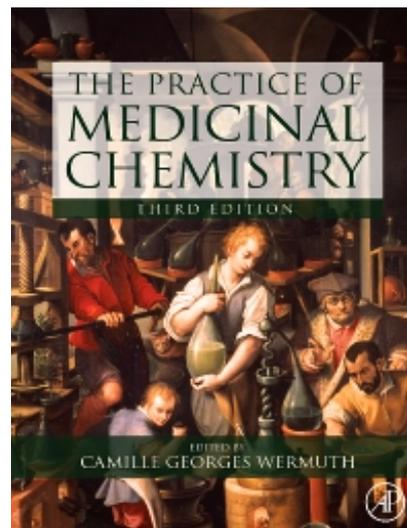
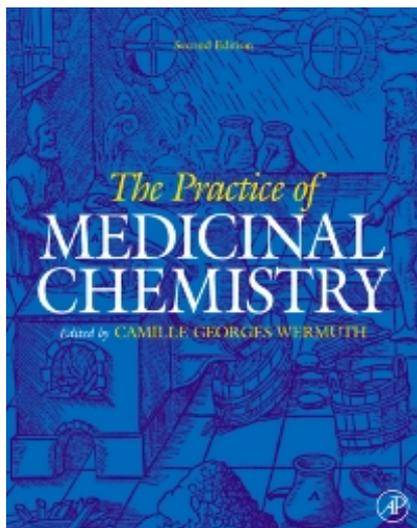
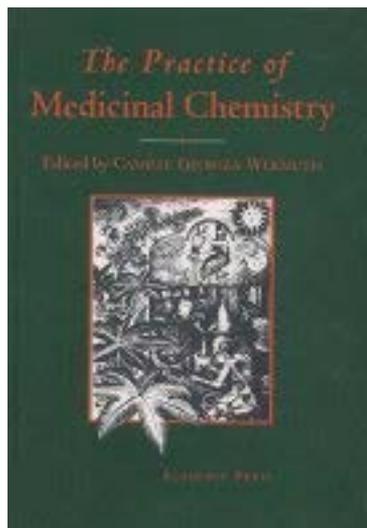


(1933-2015)

“In conclusion, the preparation of dual- or multiple-ligands on an almost rational basis is now conceivable and it can be expected that many of these molecules will yield drugs of superior clinical value compared with monotarget formulations”.

The Practice of Medicinal Chemistry

4th Edition, Elsevier, 2015.



Disclaimer. No advertisement, only proper tribute to a remarkable man and scientist.

**If we can predict by the current
chemoinformatics tools the most probable
targets for the existing drugs?**

Yes, we can!

**Both structure-based and ligand-based methods
may be applied for this purpose: (Q)SAR,
pharmacophore sets, inverse docking, etc.**

However, not all methods are freely available.

Some freely available computational tools for DRP

PASS (Prediction of Activity Spectra for Substances)

Poroikov V. et al. *Automatic Documentation and Mathematical Linguistics*, 1993, 27: 40-43.

Filimonov D. et al. *Experimental and Clinical Pharmacology*, 1995, 58: 56-62.

Lagunin A. et al. *Bioinformatics*, 2000, 16: 747-748. (www.way2drug.com/passonline)

Filimonov D. *Chemistry of Heterocyclic Compounds*, 2014, No. 3, 483-499.

SEA (Similarity Ensemble Approach)

Keiser M.J. et al. *Nat. Biotech.*, 2007, 25:197-206. (sea.bkslab.org/)

PharmMapper

Liu X. et al. *Nucl. Acids Res.*, 2010, 38, W609-W614. (59.78.96.61/pharmmapper/)

DRAR-CPI

Luo H. et al. *Nucl. Acids Res.*, 2011, 39, W492-W498. (cpi.bio-x.cn/drar/)

TargetHunter

Wang L. et al. *AAPS J.*, 2013, 15: 395-406. (www.cbligand.org/TargetHunter/)

SuperPred

Nickel J. et al. *Nucl. Acids Res.*, 2014, 42: W26-31. (prediction.charite.de/)

SwissTargetPrediction

Gfeller D. et al. *Nucl. Acids Res.*, 2014, 42, W32-W38. (www.swisstargetprediction.ch/)

ChemProt 3.0

Kringelum J. et al. *DataBase*, 2016, 2016: bav123. (potentia.cbs.dtu.dk/ChemProt/)

More info about the computational resources:



NPR

REVIEW



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Chemo- and bioinformatics resources for *in silico* drug discovery from medicinal plants beyond their traditional use: a critical review†

Cite this: *Nat. Prod. Rep.*, 2014, 31, 1585

Alexey A. Lagunin,^{*ac} Rajesh K. Goel,^{*b} Dinesh Y. Gawande,^b Priynka Pahwa,^b Tatyana A. Glorizova,^a Alexander V. Dmitriev,^a Sergey M. Ivanov,^a Anastassia V. Rudik,^a Varvara I. Konova,^a Pavel V. Pogodin,^{ac} Dmitry S. Druzhilovsky^a and Vladimir V. Poroikov^{*ac}

REVIEWS Drug Discovery Today • Volume 21, Number 1 • January 2016



ELSEVIER

Teaser In silico approaches reveal mechanisms of adverse drug reactions and predict them at the earliest stages of drug development.

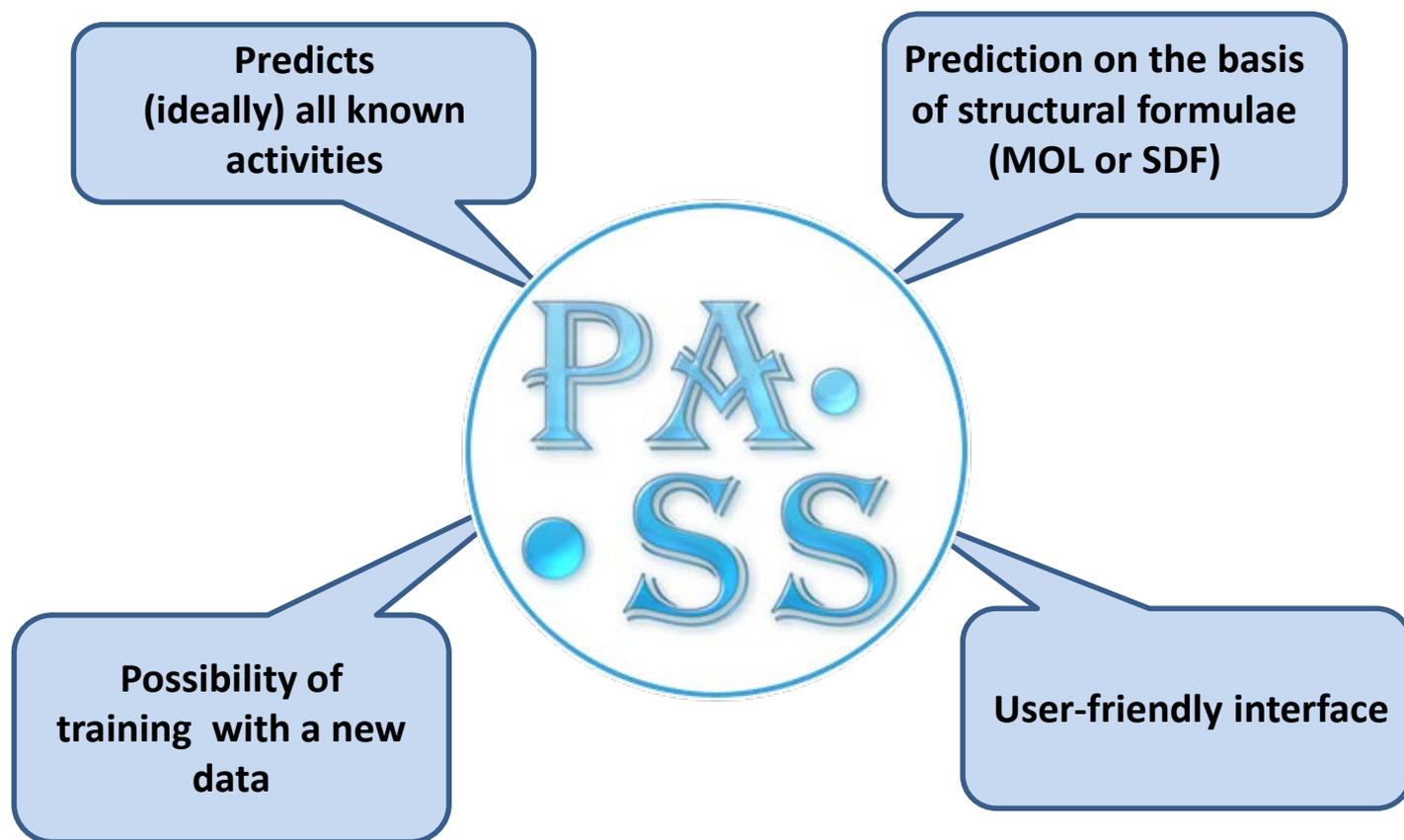
Reviews • KEYNOTE REVIEW



***In silico* assessment of adverse drug reactions and associated mechanisms**

Sergey M. Ivanov^{1,2}, Alexey A. Lagunin^{1,2} and Vladimir V. Poroikov^{1,2}

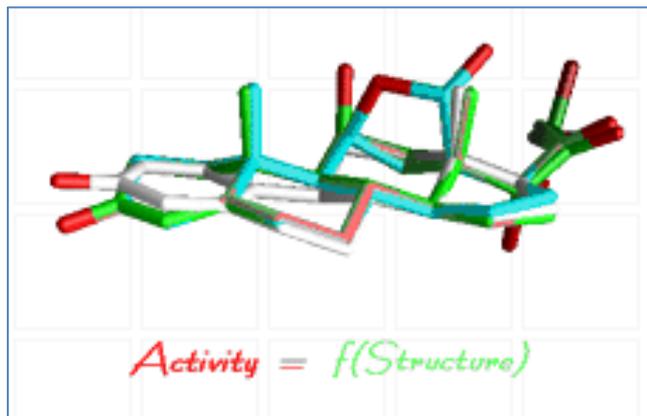
Requirements for a computer program evaluated biological activity profiles (spectra)



Biological activity spectra of organic compound

Biological activity is one of the most important characteristics of organic compound, which provides the basis for its use in therapeutic purposes. Biological activity reflects the result of interaction between the substance and biological object, and depends on substance structure and properties, biological object (species, sex, age), and mode of action (administration route, dose). Biological activity spectrum of an organic compound is the set of different kinds of biological activity that reflect the results of the compound's interaction with various biological entities. It represents the "intrinsic" property of a substance depending only on its structure. This is a qualitative characteristic property of a substance that depends only on its molecular structure.

Structure-activity relationships: (Q)SAR

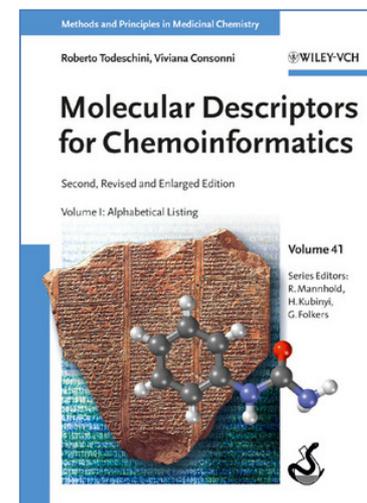


Molecular descriptors

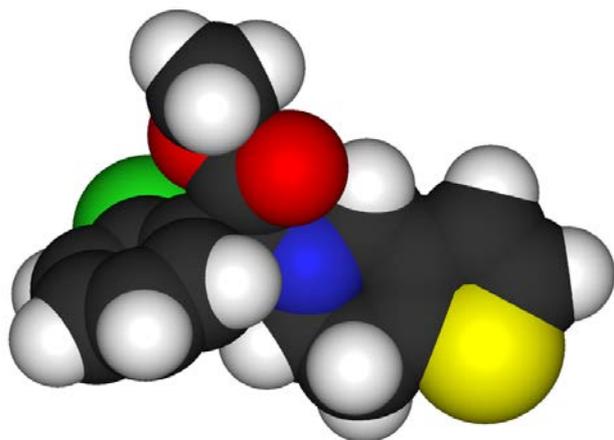
Sub-structural (-COO, -NH₂, -OH, C₆H₅, и др.); physical-chemical (molecular weight, melting point, IR frequencies, chemical shifts in NMR, etc.); molecular connectivity, Wiener indices, Balaban indices, hydrophobicity constant, pK_a, van der Waals volume, Log P, water solubility, etc. (several thousand).

Mathematical methods

Multiple linear regression (MLR); non-linear regression; partial least squares (PLS); regression on principal components (PCR); artificial neural networks (ANN); similarity matrices; genetic algorithms; support vector machine (SVM); cluster analysis (CA); discriminant analysis; etc.



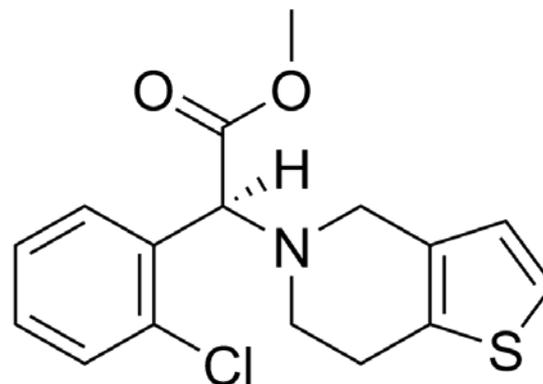
Chemical structure representation



The spatial configuration of the free uncharged molecules in the ground state in a vacuum is a necessary and sufficient description of its structure.

The use of this molecular structure description requires substantial computational resources for molecular modeling and/or quantum-chemical calculations.

However, **the basis of all calculations is the traditional structural formula.**



Thus, the structural formula uniquely determines all properties of the organic molecule.

Influence of the environment?

- **Structural formula determines, at least, potential "intrinsic" properties of the molecule.**

Neighborhoods of atoms descriptors

The most biological activities of organic compounds are the result of molecular recognition, which in turn depends on the correspondence between particular atoms of the ligand and the target.

MOLECULAR BIOLOGY

QUANTUM CHEMISTRY

QUANTUM FIELDS THEORY

$$M = V + VgM = V + VgV + VgVgV + VgVgVg + \dots$$

$$M_i = V_i + V_i g M = V_i + V_i g (M_1 + M_2 + \dots + M_m)$$



D.A. Filimonov

Descriptors are based on the concept of atoms' of molecule taking into account the influence of the neighborhoods:

MNA - **M**ultilevel **N**eighborhoods of **A**toms

QNA - **Q**uantitative **N**eighborhoods of **A**toms

LMNA - **L**abeled **M**ultilevel **N**eighborhoods of **A**toms

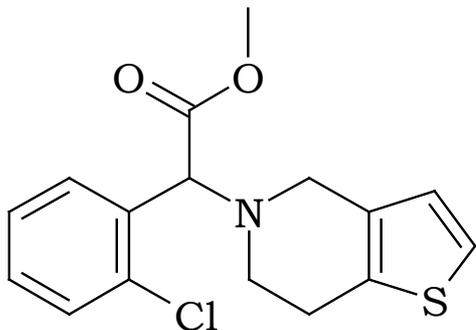
Filimonov D.A., Poroikov V.V. *In: Chemoinformatics Approaches to Virtual Screening*. Eds. Alexandre Varnek and Alexander Tropsha. Cambridge (UK): RSC Publishing, 2008, 182-216.

Filimonov D.A. et al. *SAR and QSAR Environ. Res.*, 2009, 20: 679-709.

Rudik A.V. et al. *J. Chem. Inform. Model.*, 2014, 54: 498–507.

Substance representation: Clopidogrel

Structural formula



Activity Spectrum

Abdominal pain
Acute neurologic disorders treatment
Agranulocytosis
Allergic reaction
Anaphylaxis
Anemia
Angioedema
Angiogenesis inhibitor
Antianginal
Antiarthritic
Anticoagulant
Antineoplastic
Antipsoriatic
Antithrombotic
...
112 known activities in PASS SAR Base

MNA Descriptors (1st and 2nd levels)

HC	C(C(CCC)C(CC-H-H)S(CC))
CHHHO	C(C(CCC)C(CS-H)-H(C))
CHHCC	C(C(CCC)N(CC-C)-H(C)-H(C))
CHHCN	C(C(CCS)C(CC-H)C(CN-H-H))
CHCC	C(C(CCS)C(CN-H-H)-H(C)-H(C))
CHCCN	C(C(CC-H-H)N(CC-C)-H(C)-H(C))
CHCS	C(C(CC-H)C(CC-H)-H(C))
CCCC	C(C(CC-H)C(CC-C)-H(C))
CCCS	C(C(CC-H)C(CC-C)-Cl(C))
CCCCI	C(C(CC-H)C(CC-Cl)-H(C))
CCOO	C(C(CC-H)C(CC-Cl)-C(CN-H-C))
NCCC	C(C(CC-H)S(CC)-H(C))
OC	N(C(CN-H-H)C(CN-H-H)-C(CN-H-C))
OCC	S(C(CCS)C(CS-H))
SCC	-H(C(CC-H))
CIC	-H(C(CC-H-H))
	-H(C(CN-H-H))
	-H(C(CS-H))
	-H(-C(CN-H-C))
	-H(-C(-H-H-H-O))
	-C(C(CC-C)N(CC-C)-H(-C)-C(-C-O-O))
	-C(-H(-C)-H(-C)-H(-C)-O(-C-C))
	-C(-C(CN-H-C)-O(-C)-O(-C-C))
	-O(-C(-H-H-H-O)-C(-C-O-O))
	-O(-C(-C-O-O))
	-Cl(C(CC-Cl))

PASS: Prediction of Activity Spectra for Substances

Full text publications, databases, presentations at conferences etc.

Reliable data on structure and activity of drug-like molecules

PASS Training set
(~1 mln structures)

MNA descriptors

Training
procedure

Bayesian algorithm

New molecule

SAR knowledgebase

Prediction results

Example of prediction for Ramipril

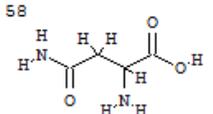
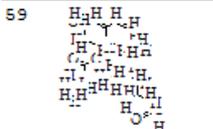
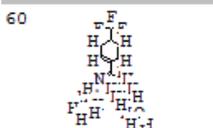
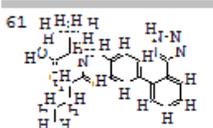
PASS - F:\DATABASES\DRUG-BANK\approved (PASS2014).SDF

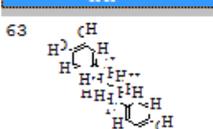
File Base Predict View Options Help

F:\PROGRAMS\PASS-2014-09-15\2014 07 20 Standard\PASS2014.SAR

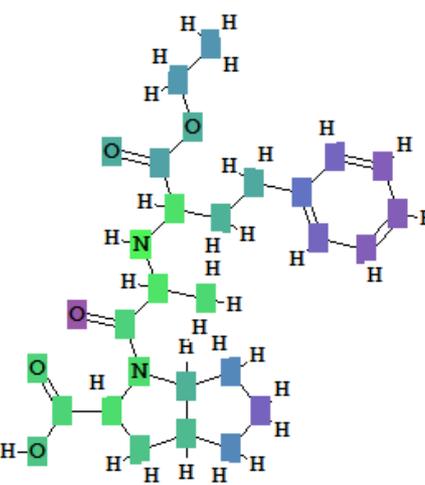
F:\DATABASES\DRUG-BANK\approved (PASS2014).SDF

5x5 4x4 3x3 2x2 Molecular Structure MNA

58

 59

 60

 61

 62

 63


> <PASS_MNA_COUNT>
 56
 > <PASS_MNA_NEW_COUNT>
 0
 > <PASS_RESULT_COUNT>
 130 of 455 Possible Pharmacological Effects
 283 of 3817 Possible Mechanisms of Action
 268 of 415 Possible Toxic and Adverse Effects



SAR Base Information

Substances	959801
Descriptors	96594
Activity Types	7598
Selected Activity Types	7157
Average IAP	0,9461
Prediction	Enabled

Effects Mechanisms Toxicity Antitargets Metabolism Gene Exp

130 of 455 Possible Pharmacological Effects at Pa > Pi

0,978	0,003	Neuroprotector
0,775	0,003	Diuretic
0,762	0,005	Vasodilator
0,665	0,006	Vasodilator, coronary
0,631	0,007	Psychostimulant
0,620	0,007	Antihypertensive
0,652	0,045	Antieczematic
0,597	0,037	Acute neurologic disorders treatment
0,568	0,018	Analeptic
0,569	0,021	Immunostimulant
0,554	0,009	Multiple sclerosis treatment
0,548	0,018	Spasmolytic
0,525	0,005	Myasthenia Gravis treatment
0,538	0,023	Antianginal
0,571	0,064	Fibrinolytic
0,518	0,058	Antineoplastic (sarcoma)
0,503	0,043	Analgesic
0,460	0,017	Antiparkinsonian
0,470	0,038	Neurodegenerative diseases treatment
0,449	0,024	Cardiotonic
0,424	0,016	Skin irritation, inactive
0,407	0,005	Saluretic
0,434	0,037	Antidiabetic
0,426	0,038	Antithrombotic
0,409	0,022	Cell adhesion molecule inhibitor
0,371	0,017	Antidiabetic symptomatic
0,391	0,040	Antimyopathies
0,392	0,046	Respiratory analgesic

56 Substructure Descriptors; 0 new.

962 of 7157 Possible Activities
 130 of 455 Possible Pharmacological Effects
 283 of 3817 Possible Mechanisms of Action
 268 of 415 Possible Toxic and Adverse Effects
 8 of 118 Possible Antitargets
 11 of 212 Possible Metabolism-Related Actions
 303 of 2298 Possible Gene Expression Regulation
 8 of 68 Possible Transporters-Related Actions

62/1411 0,620 0,007 Antihypertensive

PharmaExpert: Interpretation of the prediction results

PharmaExpert

File Tools View Help

Prediction & Interpretation - H:\DATABASES\DRUG-BANK\approved (PASS2014).SDF, 57/1278

Cholecalciferol Menadione Adenosine triphosphate L-Proline Adenine L-Asparagine Pravastatin Fluvoxamine Valsartan Rampril MasoprocolH

Save TXT Save SD Clipboard Exclude

Pa	Pi	<GENERIC_NAME>
0.930	0.001	Captopril
0.861	0.001	Fosinopril
0.852	0.001	Succimer
0.741	0.001	Dimercaprol
0.567	0.001	Spirapril
0.476	0.002	Cilazapril
0.427	0.002	Perindopril
0.401	0.002	Rampril
0.401	0.002	Trandolapril
0.354	0.002	Enalapril
0.336	0.002	Quinapril
0.314	0.002	Lisinopril
0.261	0.003	Moexipil
0.212	0.003	Glutathione
0.182	0.003	Benazepril
0.106	0.004	L-Proline
0.078	0.005	L-Isoleucine
0.095	0.007	Carlgiumic acid
0.052	0.008	Aspartame
0.049	0.008	Ibuprofen
0.047	0.009	Milgrinide
0.045	0.010	Nateglinide
0.043	0.011	Pregabalin
0.042	0.011	Lisdexamfetamine
0.039	0.014	Spironolactone
0.037	0.015	L-Arginine
0.036	0.016	Acamprosate
0.035	0.018	Dobutamine
0.035	0.018	Isoetharine
0.035	0.017	Cresatine
0.035	0.017	L-Tryptophan
0.035	0.018	Lipic Acid
0.034	0.019	Pentagastrin
0.033	0.020	Biotin
0.033	0.021	Dextroamphetamine
0.033	0.021	Amphetamine
0.032	0.023	Marimastat
0.032	0.023	Orlistat
0.032	0.021	Ticlopidimethiazide
0.032	0.023	Colistin
0.031	0.024	Alvimopan
0.031	0.024	Pravastatin
0.031	0.024	Gonadorelin
0.030	0.026	Remikiren
0.030	0.026	Saxagliptin

Number of selected compounds: 45

Pa Pi Activity Predicted value descending

Pa	Pi	Activity
0.791	0.009	Abdominal distension
0.791	0.011	Toxic, vascular
0.775	0.003	Diuretic
0.785	0.014	Excitability
0.778	0.008	Dry eye
0.782	0.017	Glucuronate 2-dehydrogenase (acceptor) inhibitor
0.763	0.002	Dermatomyositis
0.762	0.005	Vasodilator
0.725	0.016	Inflammation
0.712	0.007	Gynaecomastia
0.709	0.005	Scleroderma
0.714	0.017	Diplopia
0.707	0.013	Induration
0.714	0.023	Atrial natriuretic peptide agonist
0.710	0.019	Stevens-Johnson syndrome
0.706	0.016	Dyspnea
0.702	0.016	Swelling
0.691	0.007	QT interval prolongation
0.684	0.006	Immunotoxin
0.695	0.018	Anemia
0.701	0.026	Breast pain
0.688	0.018	Angioedema
0.671	0.005	Carcinogenic, mouse, female
0.665	0.006	Vasodilator, coronary
0.656	0.001	Henoch-Schönlein purpura
0.681	0.034	TGFβ1 expression inhibitor
0.661	0.015	Stidor
0.671	0.035	Taste disturbance
0.631	0.007	Psychostimulant
0.620	0.007	Antihypertensive

Substance intended to prevent damage to the brain or spinal cord from ischemia, stroke, convulsions, or trauma. Some must be administered before the event, but others may be effective for some time after.

Metabolism: 11 Transport: 8 Gene Expression: 303

Effect: 135 Mechanism: 283 Toxicity: 268 Antitarget: 8

Effect	Mechanism	Toxicity	Antitarget
Neuroprotector	0.978	0.003	
Diuretic	0.775	0.003	
Vasodilator	0.762	0.005	
Psychostimulant	0.631	0.007	
Antihypertensive	0.620	0.007	
Endopeptidase inhibitor	0.849	0.001	
Diuretic	0.775	0.003	
Vasodilator	0.762	0.005	
Atrial natriuretic peptide agonist	0.714	0.023	
Immunostimulant	0.569	0.021	
Saluretic	0.407	0.005	
Angiotensin-converting enzyme inhibitor	0.401	0.002	
Vasodilator, peripheral	0.232	0.161	
Uric acid excretion stimulant	0.193	0.069	
Neutral endopeptidase inhibitor	0.069	0.005	
Beta adrenoceptor antagonist	0.052	0.047	
Nitric oxide donor	0.037	0.019	
Acute neurologic disorders treatment	0.597	0.037	
Immunostimulant	0.569	0.021	
Analeptic	0.568	0.018	
Spasmolytic	0.548	0.018	
Antianginal	0.538	0.023	
Myasthenia Gravis treatment	0.525	0.005	
Analgic	0.503	0.043	
Neurodegenerative diseases treatment	0.470	0.038	
Cardiotonic	0.449	0.024	
Antidiabetic	0.434	0.037	

KEGG | NCI Pathways | Reactome | Gene Ontology

KEGG	NCI Pathways	Reactome	Gene Ontology
0.714	0.023	Atrial natriuretic peptide agonist	
		HIF-1 signaling pathway	
0.401	0.002	Angiotensin-converting enzyme inhibitor	
		Chagas disease (American trypanosomiasis)	
		Hypertrophic cardiomyopathy (HCM)	
		Renin-angiotensin system	
0.390	0.039	Interleukin 2 agonist	
		Allograft rejection	
		Autoimmune thyroid disease	
		Chagas disease (American trypanosomiasis)	
		Cytokine-cytokine receptor interaction	
		Endocytosis	
		Graft-versus-host disease	
		HTLV-1 infection	
		Intestinal immune network for IgA production	
		Jak-STAT signaling pathway	
		Measles	
		PI3K-Akt signaling pathway	
		T cell receptor signaling pathway	
		Transcriptional misregulation in cancer	
		Type I diabetes mellitus	
0.386	0.005	P-glycoprotein 1 inhibitor	
		ABC transporters	
		Bile secretion	
0.343	0.152	Insulin like growth factor 1 agonist	
		Adherens junction	
		Endocytosis	
		Focal adhesion	
		Glioma	
		HIF-1 signaling pathway	
		Long-term depression	
		Melanoma	
		Oocyte meiosis	
		Pathways in cancer	
		PKA/AK/Protein Kinase A pathway	

Therapeutic effects

- Analgic
- Analgic, non-opioid
- Antianginal
- Antialsthmatic
- Antidiabetic symptomatic
- Antidiarrheal
- Antiglaucomic
- Antihypertensive
- Antiinflammatory
- Antischismic, cerebral
- Antimigraine
- Antineoplastic
- Atherosclerosis treatment
- Cardiotonic



A. Lagunin

UniProt ID	Gene name(s)	Species

Pa > Pi No Mutagenic

Pa > Pi Antihypertensive

Pa > Pi Angiotensin-converting enzyme inhibitor

Pa = None Mutagenic

New Descriptors >= 0

Add Search

Delete Clear Load Include Save

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D. Druzhilovsky



A. Rudik



A. Zakharov

www.way2drug.com/Projects.php

Some examples of practical applications of biological activity spectra prediction



Athina Geronikaki,
AUT, Greece

Search for multitarget
drugs

J. Med. Chem. 2008, 51: 1601.

Finding molecules with
needed effects and MOA
Cur. Top. Med. Chem., 2015, 16: 441.



Marc Nicklaus,
NCI/NIH, USA

Drug repurposing
Pharm. Chem. J. 2011, 45: 605.

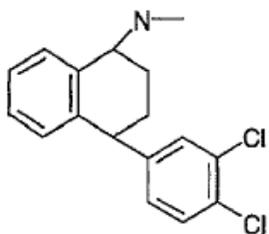


Rajesh Goel,
PU, India

Estimating drug-drug interactions
for phytoconstituents
of medicinal plants
Nat. Prod. Rep., 2014, 31: 1585.

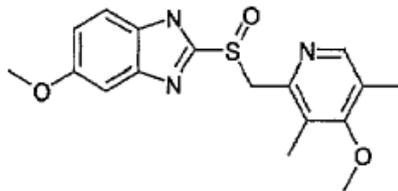


Sergey Kryzhanovsky,
Inst. of Pharmacol., Russia



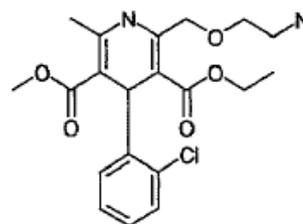
Sertraline

Adrenergic transmitter uptake inhibitor (0.770)
 Antiparkinsonian (0.609)
 Leukopoiesis inhibitor (0.582)
 Cocain dependency treatment (0.560)
 Acute neurologic disorders treatment (0.541)



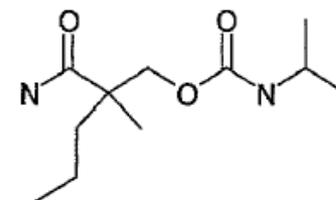
Omeprazole

TNF-alpha release inhibitor (0.658)
 Atherosclerosis treatment (0.541)



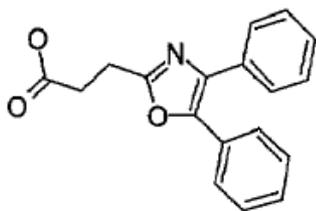
Amlodipine

Antineoplastic enhancer (0.608)
 Multiple sclerosis treatment (0.508)



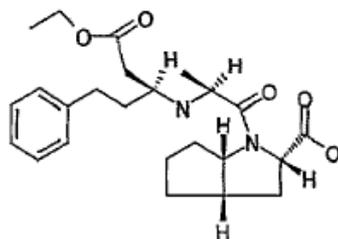
Carisoprodol

Angiogenesis inhibitor (0.569)
 Multiple sclerosis treatment (0.549)



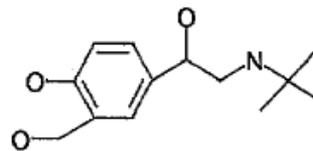
Oxaprozin

Bone formation stimulant (0.785)
 Interleukin 1 antagonist (0.640)



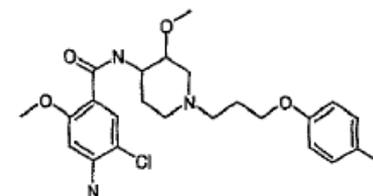
Ramipril

Multiple sclerosis treatment (0.589)
 Cognition disorders treatment (0.562)
 Antiarthritic (0.454)



Albuterol

Antiobesity (0.784)



Cisapride

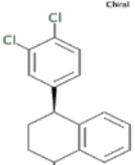
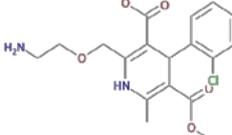
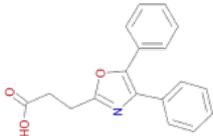
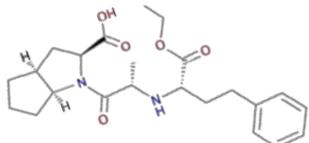
Irritable Bowel syndrome therapy (0.720)
 Rhinitis treatment (0.524)

FIGURE 3 Examples of biological activities predicted de novo for some pharmaceuticals from the Top 200 list, which may become a reason for a new application. Pa values are given in brackets.

Drug repositioning based on PASS prediction

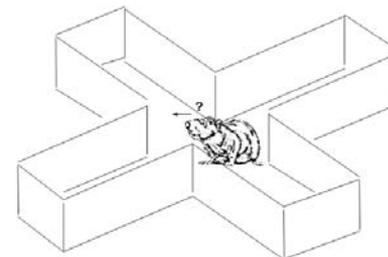
In 2001 we published predictions of new effects for 8 medicines from the list of Top200 Drugs [1].

Which predictions are confirmed? (informational search, September 2014)

	Sertraline	Cocain dependency treatment	+	Ref. [2]
	Amlodipine	Antineoplastic enhancer (moderate BCRP/ABCG2 inhibitor)	+	[3]
	Oxaprozin	Interleukin 1 antagonist (Inhibitor of production of Interleukin 1β)	+	[4]
	Ramipril	Antiarthritic	+	[5]

1. [Poroikov V. et al. SAR and QSAR Environ. Res., 2001, 12: 327-344.](#)
2. [Mancino M.J. et al. J. Clin. Psychopharmacol., 2014, 34: 234-239.](#)
3. [Takara K. et al. Mol. Med. Rep., 2012, 5: 603-609.](#)
4. [Rainsford K.D. et al. Inflammopharmacology, 2002, 10: 85-239.](#)
5. [Shi Q. et al. Arthritis Res. Ther., 2012, 14: R223.](#)

Nootropic effect in some antihypertensive drugs?



Name	Pa (Nootropic effect), %
Captopril	44,6
Enalapril	65,5
Lisinopril	61,8
Perindopril	60,9
Quinapril	65,1
Ramipril	63,3
Monopril	30,9
Piracetam	81,7
Amlodipin	-
Hydrochlorothiazide	-

Perindopril in dose of **1 mg/kg**, and **quinapril and monopril** in doses of **10 mg/kg** improved the patrolling behavior in the maze, like **piracetam and meclofenoxate** (in doses of **300 and 120 mg/kg**, respectively).

BMJ Open 2013;3:e002881 doi:10.1136/bmjopen-2013-002881

Geriatric medicine

Effects of centrally acting ACE inhibitors on the rate of cognitive decline in dementia

Yang Gao^{1,2}, Rónán O'Caomh¹, Liam Healy¹, David M Kerins^{3,4}, Joseph Eustace⁵, Gordon Guyatt⁶, David Sammon², D William Molloy^{1,7}

+ Author Affiliations

Correspondence to

Professor D William Molloy; w.molloy@ucc.ie

Published 22 July 2013

Let's validate the available computational tools for DRP

PASS (Prediction of Activity Spectra for Substances)

Poroikov V. et al. *Automatic Documentation and Mathematical Linguistics*, 1993, 27: 40-43.

Filimonov D. et al. *Experimental and Clinical Pharmacology*, 1995, 58: 56-62.

Lagunin A. et al. *Bioinformatics*, 2000, 16: 747-748. (www.way2drug.com/passonline)

Filimonov D. *Chemistry of Heterocyclic Compounds*, 2014, No. 3, 483-499.

SEA (Similarity Ensemble Approach)

Keiser M.J. et al. *Nat. Biotech.*, 2007, 25:197-206. (sea.bkslab.org/)

PharmMapper

Liu X. et al. *Nucl. Acids Res.*, 2010, 38, W609-W614. (59.78.96.61/pharmmapper/)

DRAR-CPI

Luo H. et al. *Nucl. Acids Res.*, 2011, 39, W492-W498. (cpi.bio-x.cn/drar/)

Calculation is not
finished yet
(>1 month).

TargetHunter

Wang L. et al. *AAPS J.*, 2013, 15: 395-406. (www.cbligand.org/TargetHunter/)

SuperPred

Nickel J. et al. *Nucl. Acids Res.*, 2014, 42: W26-31. (prediction.charite.de/)

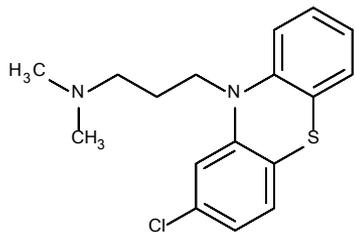
SwissTargetPrediction

Gfeller D. et al. *Nucl. Acids Res.*, 2014, 42, W32-W38. (www.swisstargetprediction.ch/)

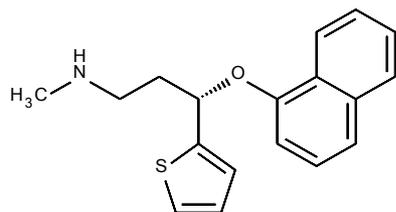
ChemProt 3.0

Kringelum J. et al. *DataBase*, 2016, 2016: bav123. (potentia.cbs.dtu.dk/ChemProt/)

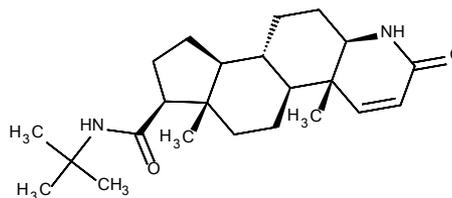
Molecules for validation of DRP computational tools



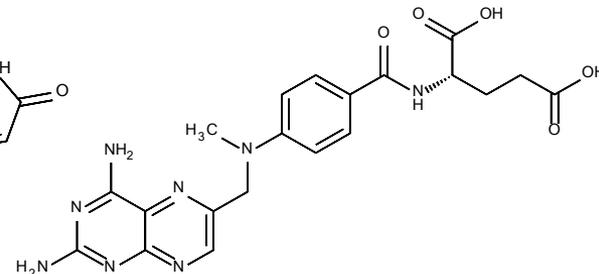
Chlorpromazine



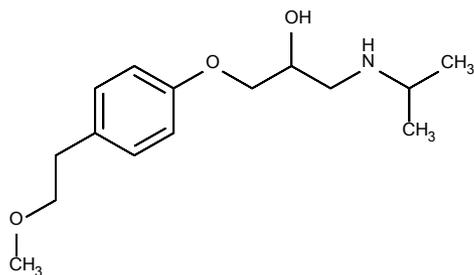
Duloxetine



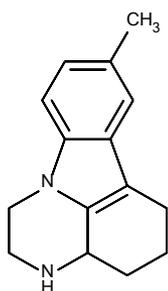
Finasteride



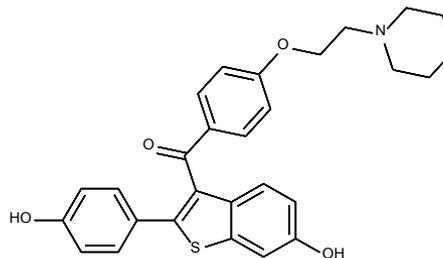
Methotrexate



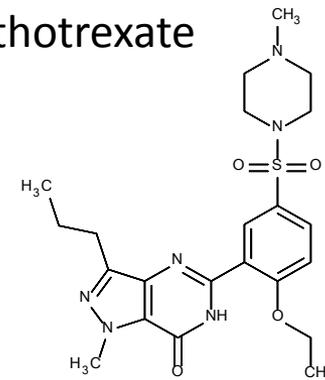
Metoprolol



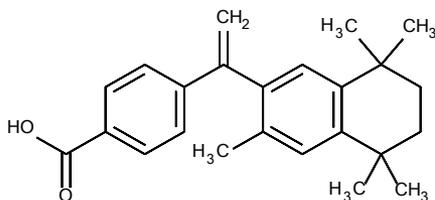
Pirlindol



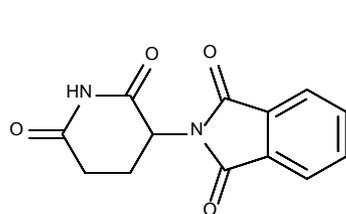
Raloxifene



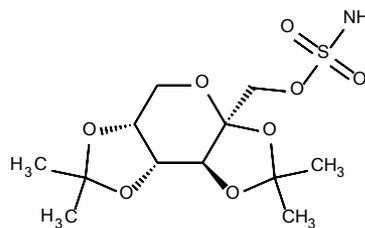
Sildenafil



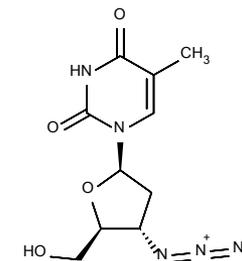
Targretin



Thalidomide



Topiramate



Zidovudine

Biological activity of compounds from validation set

Drug	Original indication	Repurposed indication
Chlorpromazine	Anti-emetic/antihistamine	Non-sedating tranquillizer
Duloxetine	Antidepressant	Stress urinary incontinence
Finasteride	Benign prostatic hyperplasia	Hair loss
Methotrexate	Acute leukemia	Osteosarcoma, breast cancer, Hodgkin lymphoma
Metoprolol	Migraine prophylaxis	Antihypertensive, Congestive heart failure
Pirlindol	Depression	Multiple sclerosis
Raloxifene	Invasive breast cancer	Osteoporosis
Sildenafil	Angina	Male erectile dysfunction
Targretin	Cancer	Alzheimer disease
Thalidomide	Sedative, nausea preventing	Leprosy, multiple myeloma
Topiramate	Epilepsy	Obesity
Zidovudine	Cancer	HIV/AIDS

Prediction of the original indications

Drug	ChP	SEA	PhM	STP	SP	TH	PASS
Chlorpromazine	+	+	--	+	+	+	+
Duloxetine	+	+	--	+	+	+	+
Finasteride	+	+	+	+	+	+	+
Methotrexate	+	+	+	+	+	+	+
Metoprolol	+	+	--	+	+	+	+
Pirlindol	+	--	+	--	--	--	+
Raloxifene	+	+	+	+	+	+	+
Sildenafil	+	+	+	+	+	+	+
Targretin	+	+	+	+	+	+	+
Thalidomide	--	--	--	--	--	--	+
Topiramate	+	--	--	--	--	--	+
Zidovudine	+	--	--	--	--	--	+

ChP – ChemProt 3.0; SEA – Similarity Ensemble Approach; PhM – PharmMapper; STP- SwissTargetPrediction; SP – SuperPred; TH – Target Hunter; PASS – Prediction of Activity Spectra for Substances.

Prediction of the repurposed indications

Drug	ChP	SEA	PhM	STP	SP	TH	PASS
Chlorpromazine	--	--	--	--	--	--	+
Duloxetine	--	--	--	--	--	--	+
Finasteride	--	--	--	--	--	--	+
Methotrexate	--	--	--	--	--	--	+
Metoprolol	+	+	--	+	+	+	+
Pirlindol	--	--	--	--	--	--	+
Raloxifene	+	+	+	+	+	+	+
Sildenafil	--	--	--	--	--	--	+
Targretin	--	--	--	--	--	--	+
Thalidomide	+	--	--	--	--	--	+
Topiramate	--	--	--	--	--	--	+
Zidovudine	+	+	--	--	--	--	+

ChP – ChemProt 3.0; SEA – Similarity Ensemble Approach; PhM – PharmMapper; STP-- SwissTargetPrediction; SP – SuperPred; TH – Target Hunter; PASS – Prediction of Activity Spectra for Substances.

Ranking of the predictions

Original indications

PharmMapper (6/12) < Target Hunter (8/12) = SuperPred (8/12) =

SwissTargetPrediction (8/12) = SEA (8/12) < ChemProt (11/12) < PASS (12/12)

Repurposed indications

PharmMapper (1/12) < Target Hunter (2/12) = SuperPred (2/12) =

SwissTargetPrediction (2/12) < SEA (3/12) < ChemProt (3/12) << PASS (12/12)

Ranking of predictions in the list (original indications)

Drug	ChP	SEA	PhM	STP	SP	TH	PASS
Chlorpromazine	209	20	--	11	11	6	66
Duloxetine	21	2	--	2	1	1	12
Finasteride	14	4	101	2	1	2	2
Methotrexate	11	1	39	5	1	2	7
Metoprolol	1	1	--	2	3	1	12
Pirlindol	1	--	19	--	--	--	2
Raloxifene	4	1	10	2	2	1	13
Sildenafil	85	1	52	2	4	2	5
Targretin	16	1	1	4	1	1	96
Thalidomide	--	--	--	--	--	--	78
Topiramate	5	--	--	--	--	--	1
Zidovudine	6	--	--	--	--	--	14

ChP – ChemProt 3.0; SEA – Similarity Ensemble Approach; PhM – PharmMapper; STP- SwissTargetPrediction; SP – SuperPred; TH – Target Hunter; PASS – Prediction of Activity Spectra for Substances.

Ranking of predictions in the list (repurposed indications)

Drug	ChP	SEA	PhM	STP	SP	TH	PASS
Chlorpromazine	--	--	--	--	--	--	105
Duloxetine	--	--	--	--	--	--	20
Finasteride	--	--	--	--	--	--	105
Methotrexate	--	--	--	--	--	--	19
Metoprolol	1	1		3	1	1	12
Pirlindol	--	--	--	--	--	--	1
Raloxifene	4	1	10	2	1	1	6
Sildenafil	--	--	--	--	--	--	3
Targretin	--	--	--	--	--	--	1095
Thalidomide	3	--	--	--	--	--	8
Topiramate	--	--	--	--	--	--	26
Zidovudine	34	2	--	--	--	--	19

ChP – ChemProt 3.0; SEA – Similarity Ensemble Approach; PhM – PharmMapper; STP-- SwissTargetPrediction; SP – SuperPred; TH – Target Hunter; PASS – Prediction of Activity Spectra for Substances.

Ranking of the predictions (taking into account the positions in the list)

Original indications

Target Hunter < SuperPred < SwissTargetPrediction < SEA < PharmMapper < **PASS**
< **ChemProt**

Repurposed indications

Target Hunter = SuperPred < SEA < SwissTargetPrediction < PharmMapper <
ChemProt < **PASS**

Clopidogrel: predicted vs. known activities

Abdominal pain	Conjunctivitis	Henoch-Schonlein purpura	Purinergic P2 antagonist
Acute neurologic disorders treatment	Consciousness alteration	Hepatic failure	Purinergic P2T antagonist
Agranulocytosis	Constipation	Hepatitis	Purinergic P2Y antagonist
Allergic reaction	Cough	Hepatotoxic	Purinergic P2Y12 antagonist
Anaphylaxis	CYP2 substrate	Hypertensive	Purinergic receptor antagonist
Anemia	CYP2C substrate	Hyperthermic	Purpura
Angioedema	CYP2C19 inhibitor	Hypotension	Renal colic
Angiogenesis inhibitor	CYP2C19 substrate	Infection	Reproductive dysfunction
Antianginal	CYP2C9 inhibitor	Insomnia	Rhinitis
Antiarthritic	CYP3A substrate	Lassitude	Sensory disturbance
Anticoagulant	CYP3A4 substrate	Leukopenia	Serum sickness
Antineoplastic	Cytochrome P450 inhibitor	Lichen planus	Shock
Antipsoriatic	Dermatitis	Lichenoid eruption	Sinusitis
Antithrombotic	Dermatologic	Malaise	Sleep disturbance
Anxiety	Dizziness	Menstruation disturbance	Stomatitis
Arthralgia	Drug eruption	Myalgia	Syncope
Atherosclerosis treatment	Dyspepsia	Nausea	THBS1 expression enhancer
Back pain	Emetic	Necrosis	Thrombocytopenia
Behavioral disturbance	Eosinophilia	Nephrotoxic	Toxic
Blindness	Erythema	Neuroprotect	Toxic epidermal necrolysis
Bronchoconstrictor	Erythema multiforme	Neutropenia	Toxic, gastrointestinal
Cardiotoxic	Exanthema	Ocular toxicity	TP53 expression enhancer
Cataract	Flatulence	Pain	Urticaria
CCL4 expression enhancer	GP IIb/IIIa receptor antagonist	Pancreatitis	Vasculitis
CCL5 expression enhancer	Hallucinogen	Pancytopenia	Vertigo
Chest pain	Headache	Platelet aggregation inhibitor	Vision disturbance
Colic	Heart failure	Platelet antagonist	
Colitis	Hematotoxic	Pruritus	
	Hemorrhage	Pulmonary embolism	

Blue – predictions coincided with the experiment.

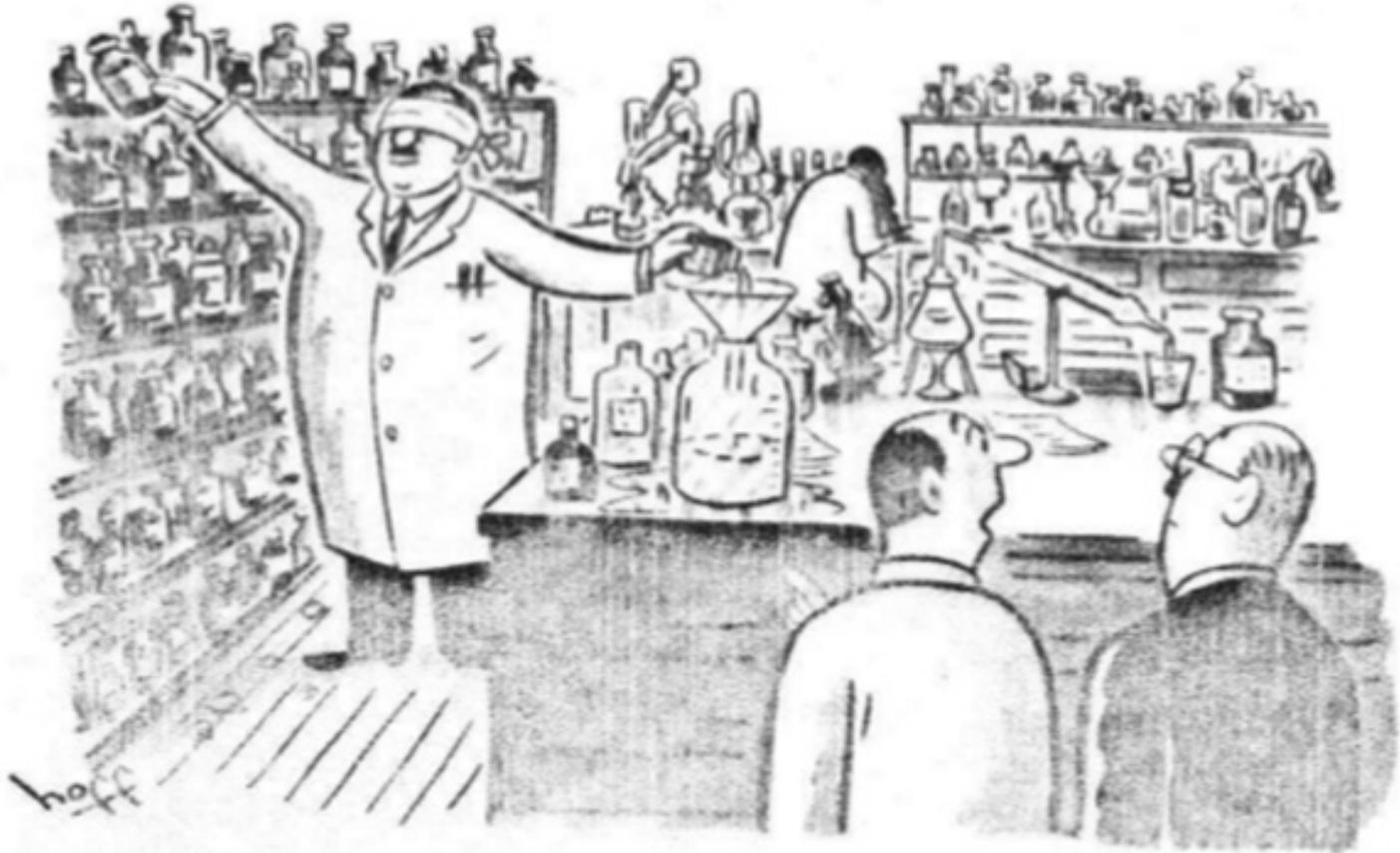
Black – unpredictable activities.

Red – unpredicted activities.



“Not all repositioning projects that work on paper are really feasible,” says Tudor Oprea, a bioinformatics researcher at the University of New Mexico in Albuquerque who monitors the field in addition to doing his own repositioning work. For instance, he says, **side effects that would be acceptable for a life-threatening disease might not be acceptable for a chronic one.** And the standard business case for repositioning — that **costs are slashed because safety tests are already in the bag — works only if the dose and mode of administration remain similar.** If the new disease requires a significantly higher dose, the drug will have to go through phase I trials again. In the end, says Oprea, development costs can be similar to those for a new molecule”.

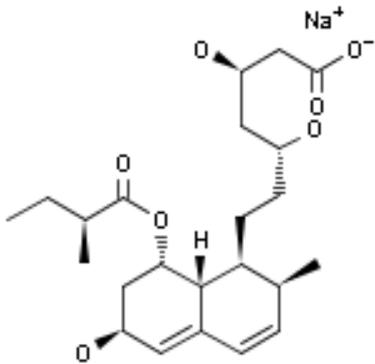
Let me remind you that our knowledges in life sciences
are rather incomplete



'That's Dr Arnold Moore. He's conducting an experiment to test the theory that most great scientific discoveries were hit on by accident.'

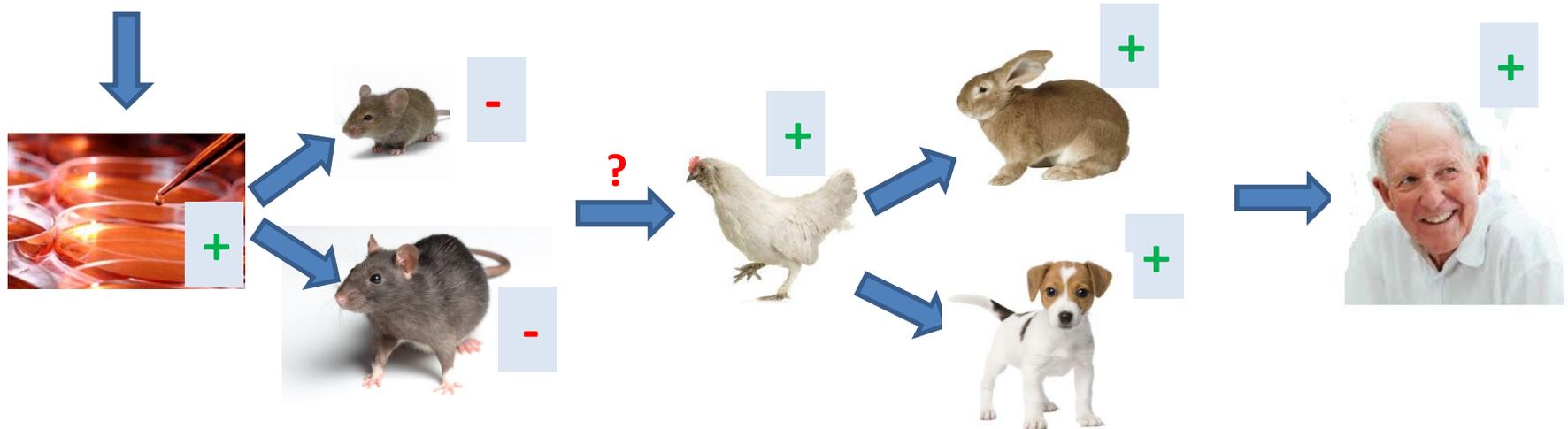
*Drawing by Hoff ; © 1957
The New Yorker Magazine, Inc.*

The history of Pravastatin development by Sankyo



HMG-CoA
reductase inhibitor

CS-514, pravastatin - derivative ML236B (compactin), which was extracted from fungus *Penicillium citrinum* in 1970 by Sankyo Pharma Inc. In 1989 Pravastatin sodium was registered as hydroxymethylglutaril-CoA-reductase inhibitor for treatment of familial hypercholesterolemia and hyperlipidemia. In 2005 Pravachol (Pravastatin sodium) became blockbuster in US with annual sales 1,3 billion dollars.



[Diabetes Res Clin Pract.](#) 1986 Jun;2(3):179-81.

Effect of CS-514, an inhibitor of 3-hydroxy-3-methylglutaryl coenzyme A reductase, on lipoprotein and apolipoprotein in plasma of hypercholesterolemic diabetics.

Yoshino G, Kazumi T, Kasama T, Iwatani I, Iwai M, Inui A, Otsuki M, Baba S.



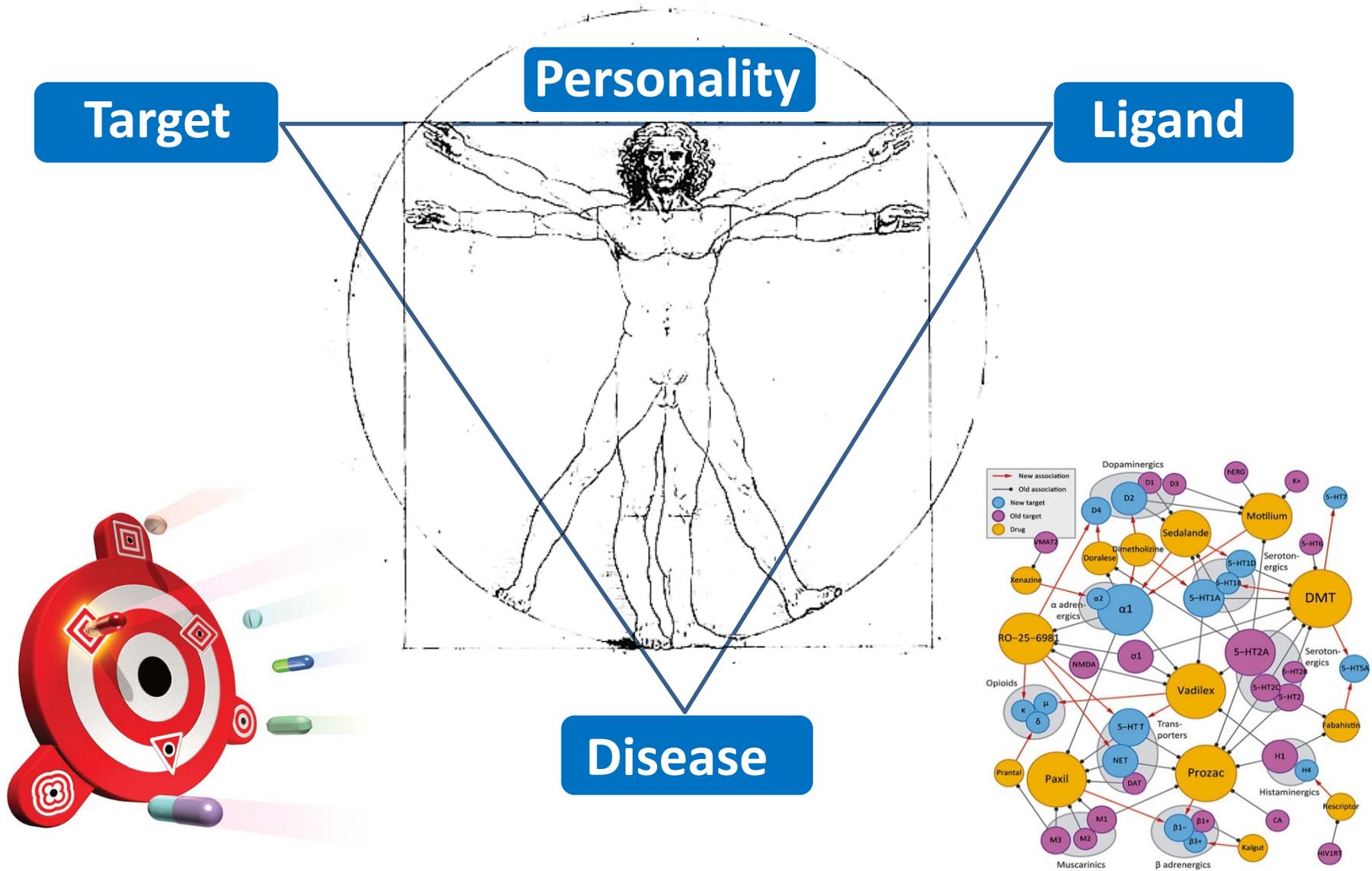
Ivan Pavlov



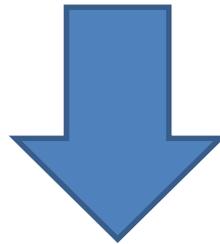
The Nobel Prize in Physiology or Medicine 1904 was awarded to Ivan Pavlov "in recognition of his work on the physiology of digestion, through which knowledge on vital aspects of the subject has been transformed and enlarged".

“On the vast territory of medical knowledge pharmacology seems, one may say the border, where there is a particularly lively exchange of services between the natural scientific basis of medicine, physiology, and medical knowledge - therapy, and where therefore particularly felt the mutual usefulness of one knowledge to another. Pharmacology, studying animal drug action by using physiological methods, improving therapy, puts it on a rational solid ground; on the other hand, the treatment indication, subjected to laboratory analysis, often leads to the discovery of the such physiological phenomena that would remain undetected for a long time with pure physiological study.”

No matter, where you start from...



**Drug Repurposing: New Uses for Old Drugs
or Systems Biomedicine?**



**Drug Repurposing: New Uses for Old Drugs
and Systems Biomedicine.**

Summary

- **Drug repurposing is a promising way for finding new medicines.**
- **“Not all repositioning projects that work on paper are really feasible” (T. Oprea).**
- **Chemoinformatics methods help to identify the most prospective directions of research.**
- **There are still some “rooms”, to improve the existing and develop novel computational methods for DRP.**
- **Drug repurposing provides opportunities for both finding new uses of old drugs and development of the systems biomedicine.**

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RESEARCH & INNOVATION
FP7



МИНИСТЕРСТВО ОБРАЗОВАНИЯ И НАУКИ
РОССИЙСКОЙ ФЕДЕРАЦИИ

РОССИЙСКИЙ НАУЧНЫЙ ФОНД
ПОДДЕРЖКА И РАЗВИТИЕ



Thank you for your kind attention!



We are open for collaboration.

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