

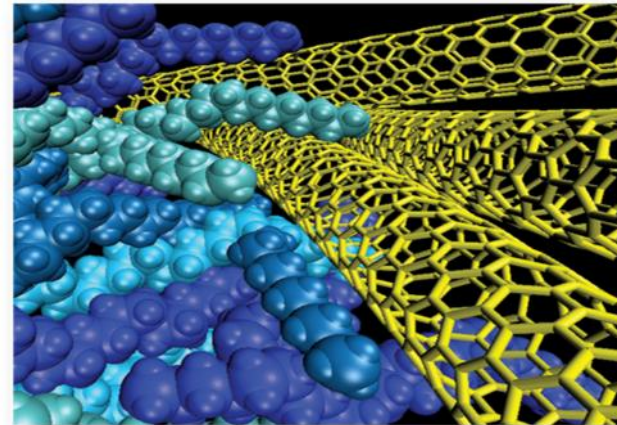
Data relevance in pharmaceutical industry

Davide Branduardi
Applications Scientist
Schrödinger, Inc.
London, UK

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What does Schrödinger do?

- Mission
Improving human health and quality of life through *advanced computational methods*
- Provides integrated software solutions and services to pharmaceutical/biotechnology and materials companies



Who is Schrödinger?

- **Founders – Scientists from Academia**
 - Richard Friesner – Columbia University
 - Theoretical chemist focused on life sciences
 - Bill Goddard – Caltech
 - Theoretical Chemist focused on materials science
- **Investors – Patient; passionate about science**
 - David E. Shaw
 - Founder of D.E. Shaw Group, Hedge Fund
 - Chief Scientist – D. E. Shaw Research
 - Senior Research Fellow – Center for Computational Biology and Bioinformatics at Columbia University
 - Bill Gates
 - No institutional investors

Schrödinger **Offices** and **Business Partners**



Schrödinger contribution to structure-based drug discovery

Scientific advances in drug discovery; for example:

- **2004**: Glide – de facto standard in protein ligand docking
- **2005**: 1st reliable flexible-receptor ligand docking method (induced fit)
- **2009**: 1st rigorous treatment of protein desolvation ('hydrophobic effect')
- **2011**: Most accurate small-molecule force field
- **2014**: 1st benchmark method for accurate prediction of binding affinity

...together with a commitment in the open source visualization software Pymol.

Some Facts & Figures

- 24 Years of innovations in scientific research and product development
- ~350 employees, >55% Ph.D.
 - Scientists
 - Engineers
- Significant R&D effort and focus on customer support
 - R&D spending: ~50% of budget
 - Development: ~50% of employees
 - Internal Drug Discovery: ~10% of employees
 - Customer Support: ~15% of employees
- Revenue is reinvested in research and development
- Focus on discovery software & services for small molecules, biologics, and materials science
- Customers: 380 commercial (including all top 30 Pharma companies); 2100 academic; 130 government

Nimbus Therapeutics

- Nimbus is pioneering a new computational technology-driven paradigm to rapidly advance a diverse pipeline into clinical development
- \$72 Million from 7 Investors
 - Including Atlas Venture, Bill Gates, and Pfizer Ventures
- Schrödinger is a founding partner (please refer to www.nimbustx.com for the up-to-date information)

TARGET		COMP. CHEM.	LEAD OPT.	IND ENABLING	PH. 1	PH. 2
ACC	Nonalcoholic Steatohepatitis (NASH) NDI-010976	▶				
	Hepatocellular Carcinoma (HCC)	▶				
IRAK4	Oncology, Immunology	▶				
Tyk2	Immunology, Oncology	▶				
KRas	Solid Tumors, Others	▶				
Additional Pipeline Targets		▶				
MONSANTO	Novel broad-spectrum fungicides	▶ Crop Field Trials				
Shire	Lysosomal Storage Disorders	▶				





- Overview
- Press Releases
- Publications & Abstracts

GILEAD SCIENCES ANNOUNCES ACQUISITION OF NIMBUS THERAPEUTICS' ACETYL-COA CARBOXYLASE (ACC) PROGRAM FOR NASH AND OTHER LIVER DISEASES

FOSTER CITY, Calif. & CAMBRIDGE, Mass., April 4, 2016—Gilead Sciences, Inc. (NASDAQ: GILD) and Nimbus Therapeutics, LLC today announced that the companies have signed a definitive agreement under which Gilead will acquire Nimbus Apollo, Inc., a wholly-owned subsidiary of Nimbus Therapeutics. Nimbus Therapeutics will receive an upfront payment of \$400 million and up to \$800 million in development milestones over time.

[Join Us For An Exclusive Webcast on Social Security](#)



News

April 4th, 2016

Schrödinger congratulates its collaboration partner, Nimbus Therapeutics, on the acquisition by Gilead of Nimbus's ACC program for up to \$1.2 billion

Gilead Sciences, Inc. (NASDAQ:GILD) and Nimbus Therapeutics, LLC, announced that the companies have signed a definitive agreement under which Gilead will acquire Nimbus Therapeutics' Acetyl-CoA Carboxylase (ACC) inhibitor program for up to \$1.2 billion.

Schrödinger is a co-founder and equity holder of Nimbus and provided the computational platform and support for the ACC program. For more information, please read the press release

Forbes / Pharma & Healthcare

APR 4, 2016 @ 08:33 AM 5,191 VIEWS

Atlas-Backed Nimbus Delivers Its Apollo Mission: A \$1.2B Gilead Partnership



Bruce Booth, CONTRIBUTOR

I write about emerging life-science startups and the industry.

FOLLOW ON FORBES (281)



Opinions expressed by Forbes Contributors are their own.

FULL BIO

Today Gilead announced the acquisition of Nimbus' 1 program targeting NASH and related metabolic disorder. The deal includes a \$400M upfront and up to another \$800M in development regulatory milestones (here). This transformative deal is the culmination of over five years of work on the program.



Biotech

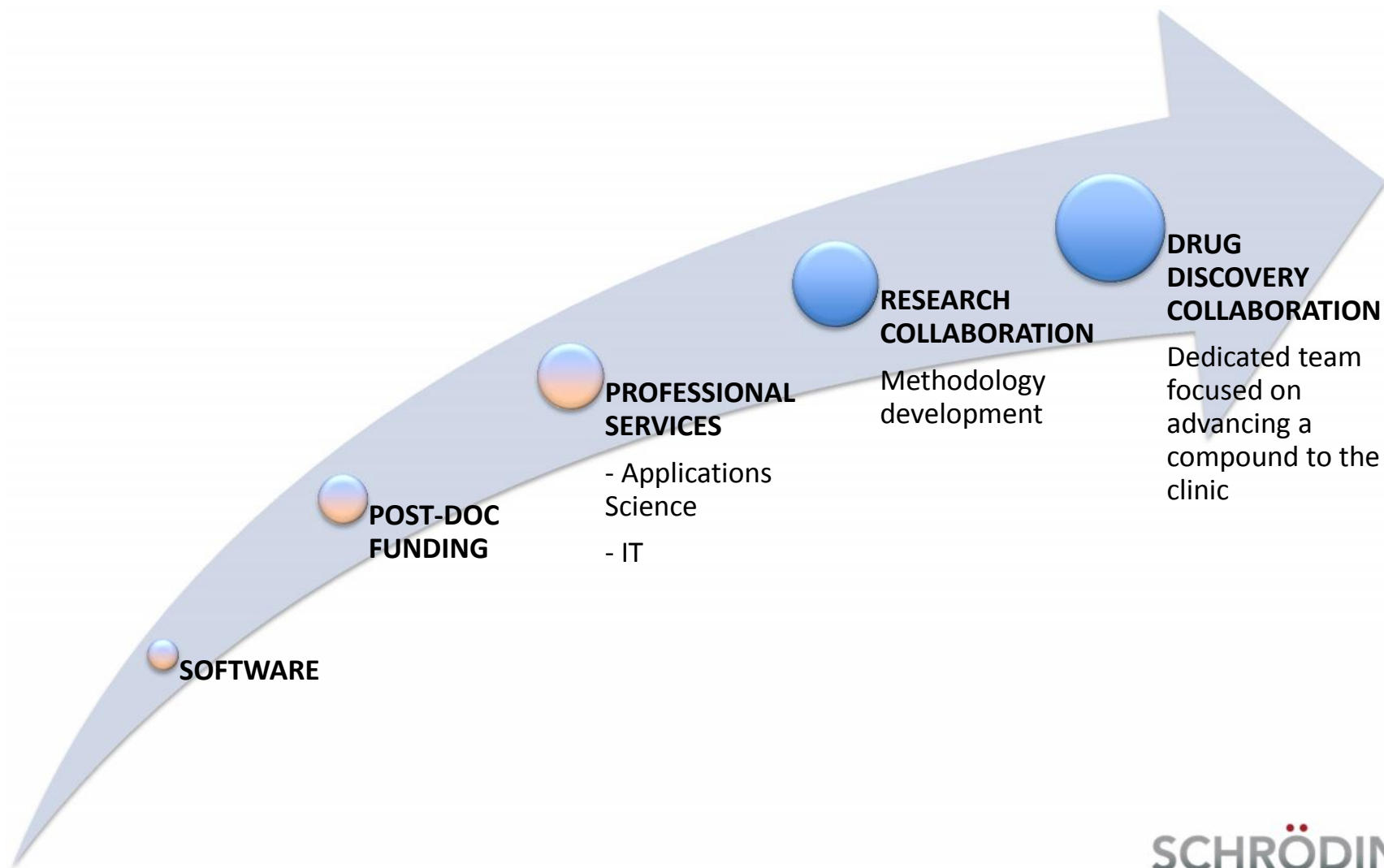
Gilead pays Nimbus \$200M after fast progress in NASH

by Nick Paul Taylor Nov 2, 2016 6:24am

Six months after Gilead paid Nimbus Therapeutics \$400 million to buy a NASH drug, it has handed over another \$200 million milestone payment.



Schrödinger works operates in many ways



Outline

- How a drug works and how is identified
- Pharma industry and data generation
 - What kind of data pharma industry generates
 - R&D issues: integration data challenge
 - Productivity
- Smart use of in-house data
- Smart use of external data
- A look to the future

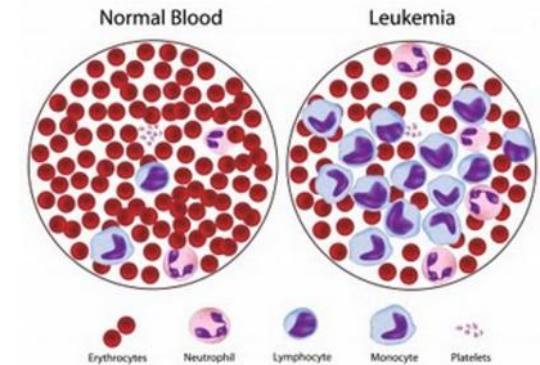
Data in Pharma

- Pharma is an interesting example of data science
 - **Research data** on drugs is very private. Attempts and failures are kept hidden for competition and driving stock price. But acceptance from the specialists happen on public
 - **Production data** is very open: regulatory agencies may want companies track batch numbers and difformities (e.g. In 2012-2013 flu pandemic, production was not effective for a change in production standard or Quinavaxem on hold in 2010)

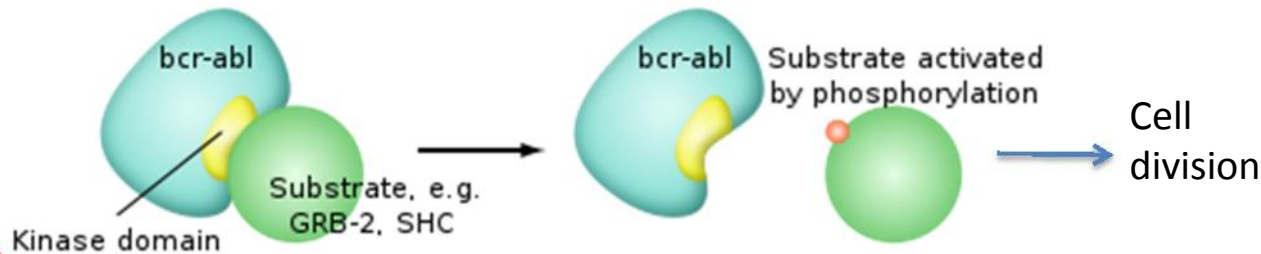
http://www.who.int/immunization_standards/vaccine_quality/outcome_quinvaxem_investigation_february_2011/en/

How a drug works?

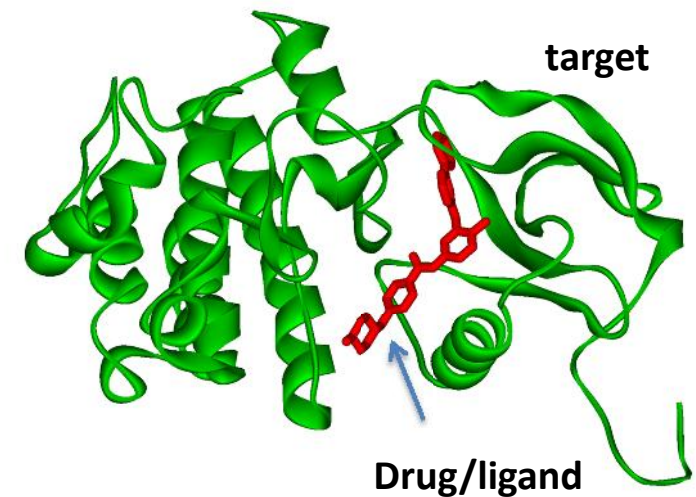
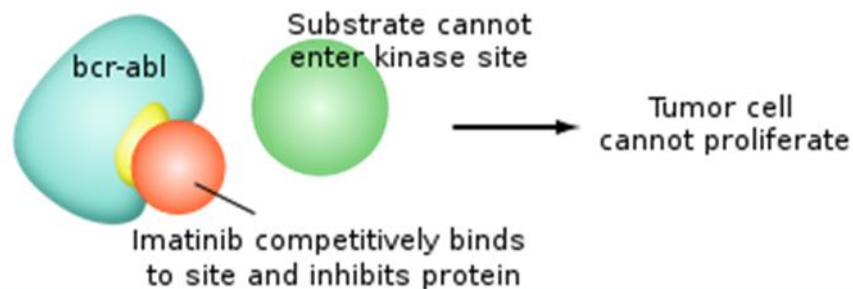
- An example: Chronic Myeloid Leukemia
- We haven't built our human body: finding mechanism (pathways) is hard



Tumor cell has aberrant replication without reaching maturation



“Inhibition” is a strategy where a chemical interrupt a “pathway”



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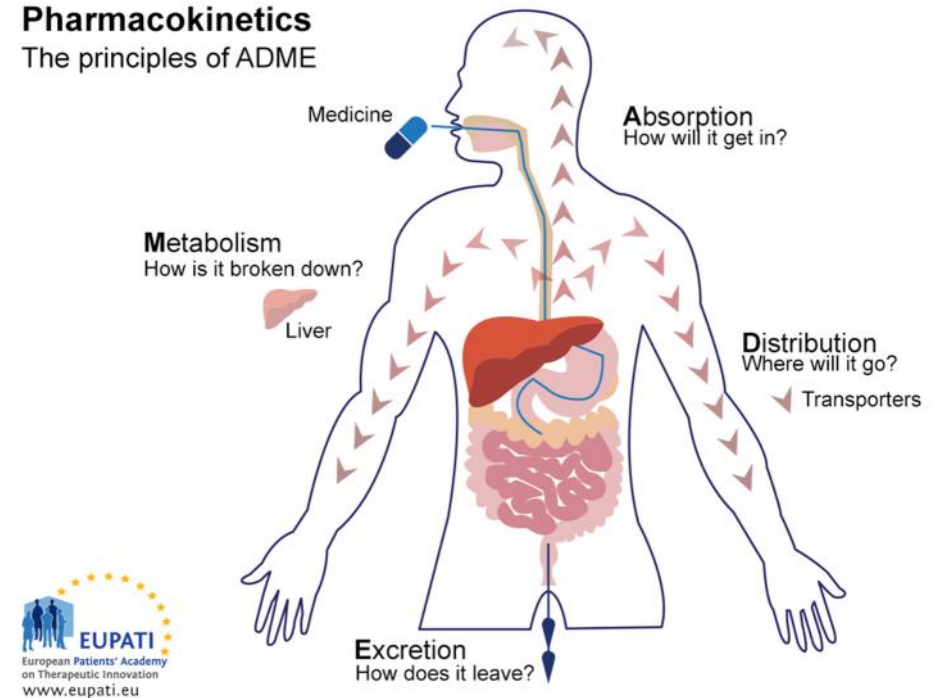
How a drug works? From your mouth to the cell

- Lots of things may happen from your mouth to the cell
 - May not penetrate the gut
 - May not be transported efficiently by blood
 - May be cleansed by liver very fast (not around for enough time to be effective)
 - May get more than anything to something which is not its targeted (TOX!)
 - Metabolites can be cleared very fast

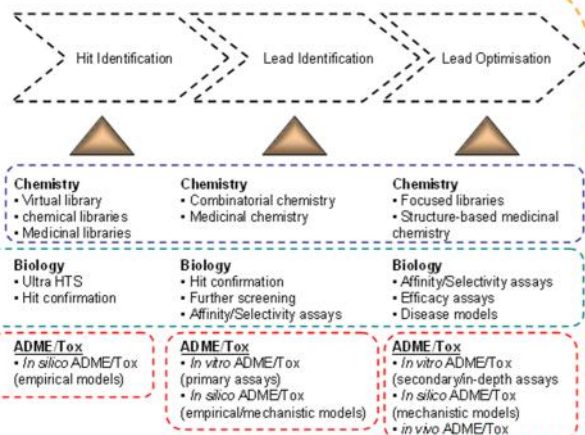
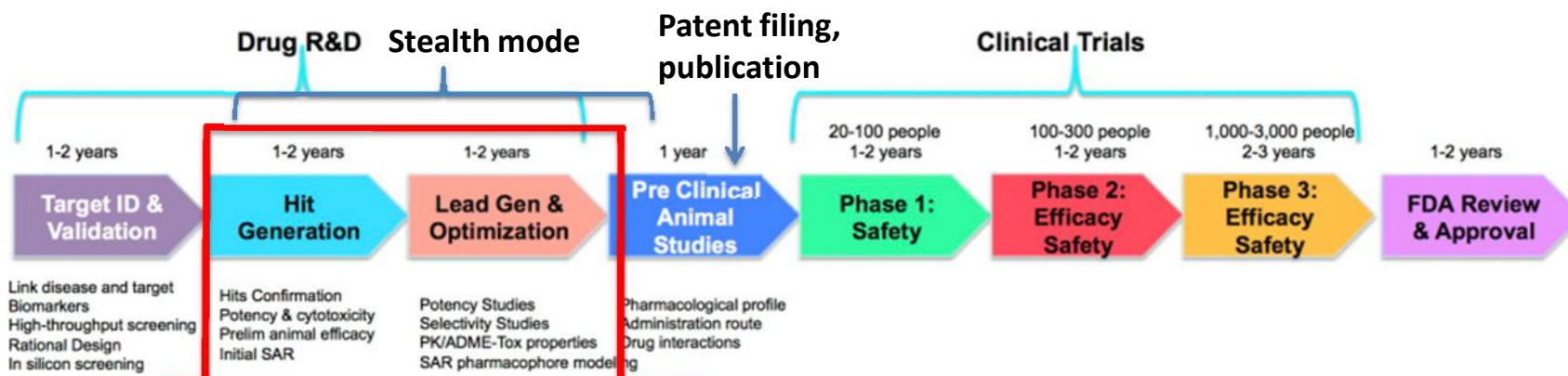


We do not know all about how our body works. We use animal studies to get as close as possible to the real scenario and also here it often does not work!

Pharmacokinetics The principles of ADME



How drugs are identified?



Molecular modelling here!

High Throughput screening robot.
10⁵ compounds screened in weeks



Libraries: millions of compounds



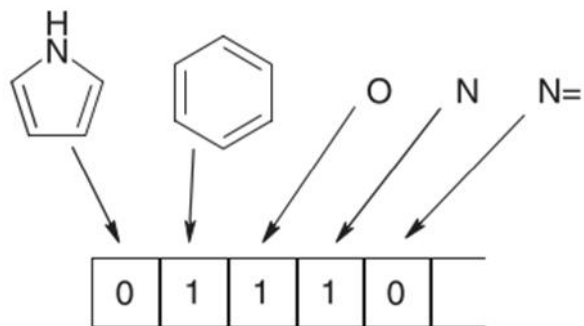
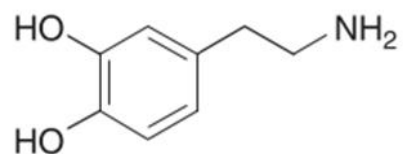
<http://www.frost.com/prod/servlet/market-insight-print.pag?docid=135570876>

www.brooks.com

<https://newdrugapprovals.org/2014/02/>

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Library design basic concepts



Bitstring (or fingerprints)



A

1	0	1	1	1	0	1	1	0	0	1	1
---	---	---	---	---	---	---	---	---	---	---	---

 $a=8$

B

0	0	1	1	0	0	1	0	1	0	1	1
---	---	---	---	---	---	---	---	---	---	---	---

 $b=6$

$c=5$

$$S_{AB} = \frac{5}{8+6-5} = 0.56$$

Tanimoto similarity



Clustering



Filtering

Prioritize compounds with Molecular modeling: in-silico approaches



3D structures of targets



Virtual Database of compounds, filtered for the purpose



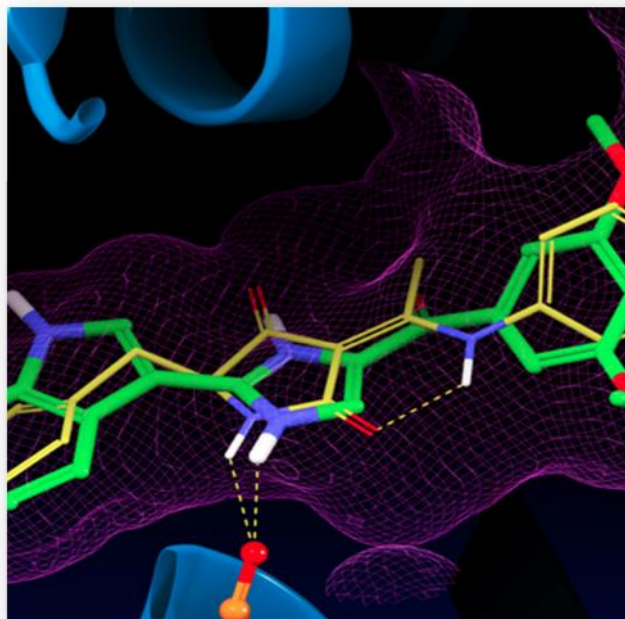
Ideas



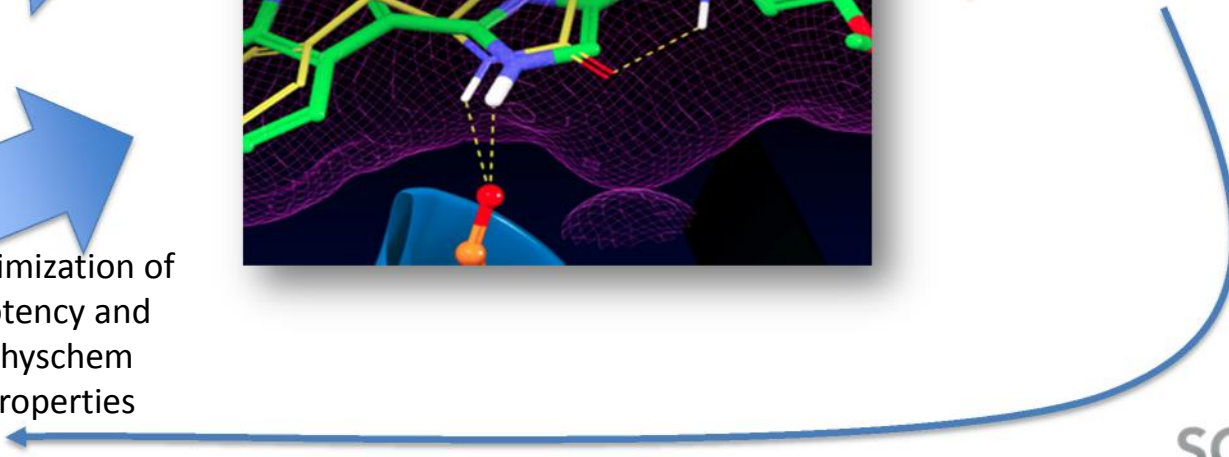
Optimization of potency and physchem properties



Rigid docking/Free Energy Perturbation (via MD simulations)

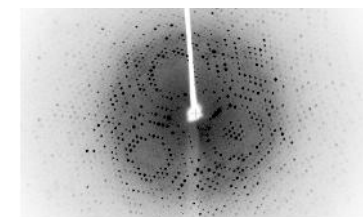
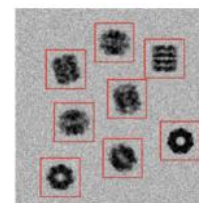
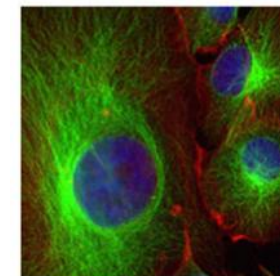
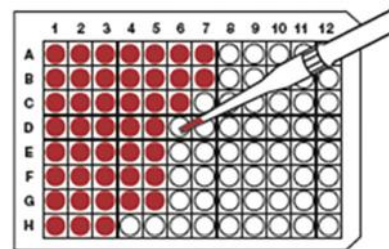


Ranking, Rationalize, Water network analysis

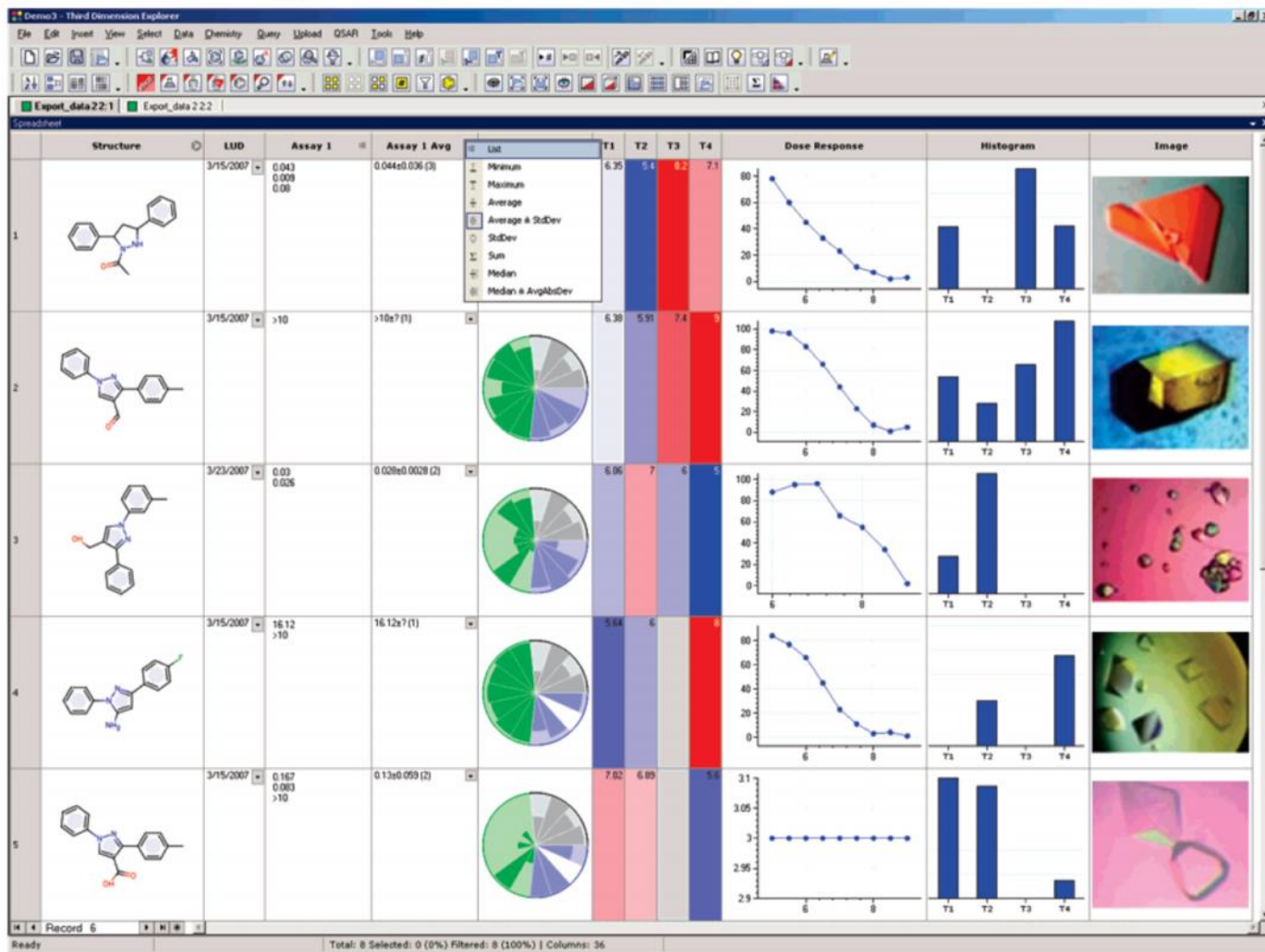


Data in R&D is everywhere and very heterogeneous

- Managing compounds in stock (availability, characterization, planning, production): 10^6
- Managing assay data
 - multiple experimental sources
 - Images
 - numbers
 - multiple reagents
 - multiple operators
- Managing structural data
 - Xray crystallography
 - Cryo-EM
 - NMR
 - Molecular modeling results
- All these data can be non integrated and redundant/outdated
- Data integration and analytics on all these
 - Spotfire (TIBCO)
 - D360 (Certara)
 - LiveDesign (Schrodinger)
- Managing Electronic lab nootebooks for intellectual property issues



Example of data integration: Janssen ABCD



Agrafiotis et. al *J. Chem. Inf. Model.*, Vol. 47, No. 6, 2007 2009

Integrating experiments and calculations: ideation engine



- LiveDesign™ is a browser-based enterprise platform
- Centralizes your small molecule data, ideas, and communication
- Designed to improve project efficiency

SAR
exploration

The screenshot displays the LiveDesign software interface. On the left, a sidebar contains navigation options like "Compounds", "Data & Columns", "Filter", "SAR Analysis", and "Viewer". The main area is divided into two panels. The top panel shows a table of compounds with columns for "Compound Structure", "ID", "Quick Properties (MW)", "Quick Properties (clogP)", and "Quick Properties (PSA)". The bottom panel shows a SAR exploration table with columns for "Compound Structure", "ID", "R1 (SAR)", "R2 (SAR) -1", "R3 (SAR)", and "Scaffold Name".

Compound Structure	ID	Quick Properties (MW)	Quick Properties (clogP)	Quick Properties (PSA)
	CRA-035000	389.88	4	145.01
	V35386	396.47	4	134.38
	V35387	408.5	4	159.08

Compound Structure	ID	R1 (SAR)	R2 (SAR) -1	R3 (SAR)	Scaffold Name
	CRA-035000				Scaffold 1
	V35386				Scaffold 1
	V35387				Scaffold 1

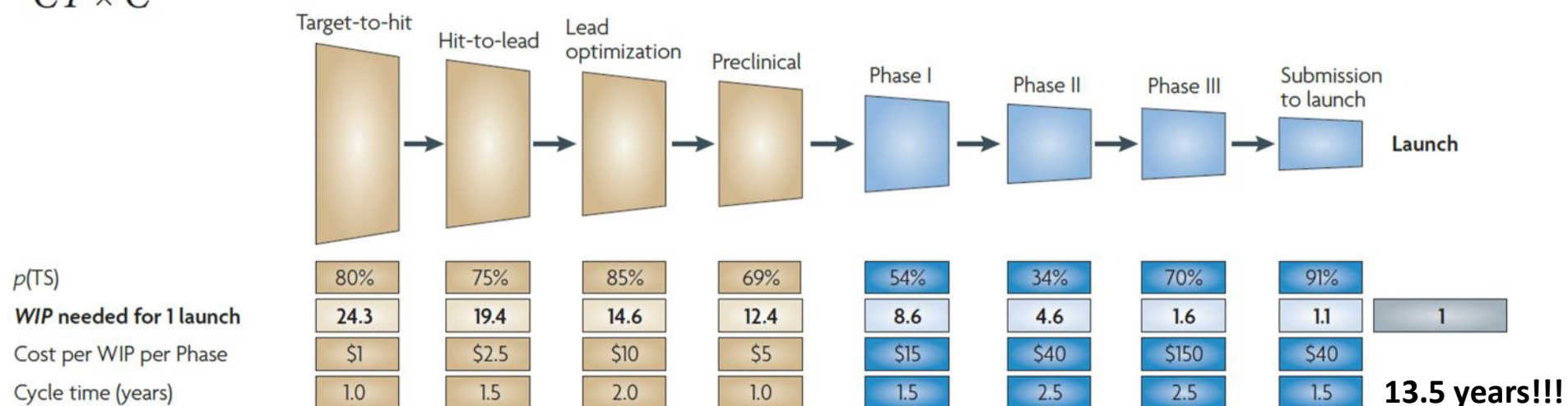
On the right, a "3D 3D Visualizer" window shows a cyan ribbon representation of a protein structure with a small molecule ligand bound in its active site. A blue arrow points from the SAR exploration table to this 3D visualization. Below it, another "3D 3D Visualizer" window shows a more detailed view of the ligand-protein interaction, with atoms represented by colored spheres and bonds by sticks. A blue arrow points from the text "3D Visualization and modeling" to this window.

3D Visualization
and modeling

From ideation to market: the path of a drug

$$P \propto \frac{WIP \times p(TS) \times V}{CT \times C}$$

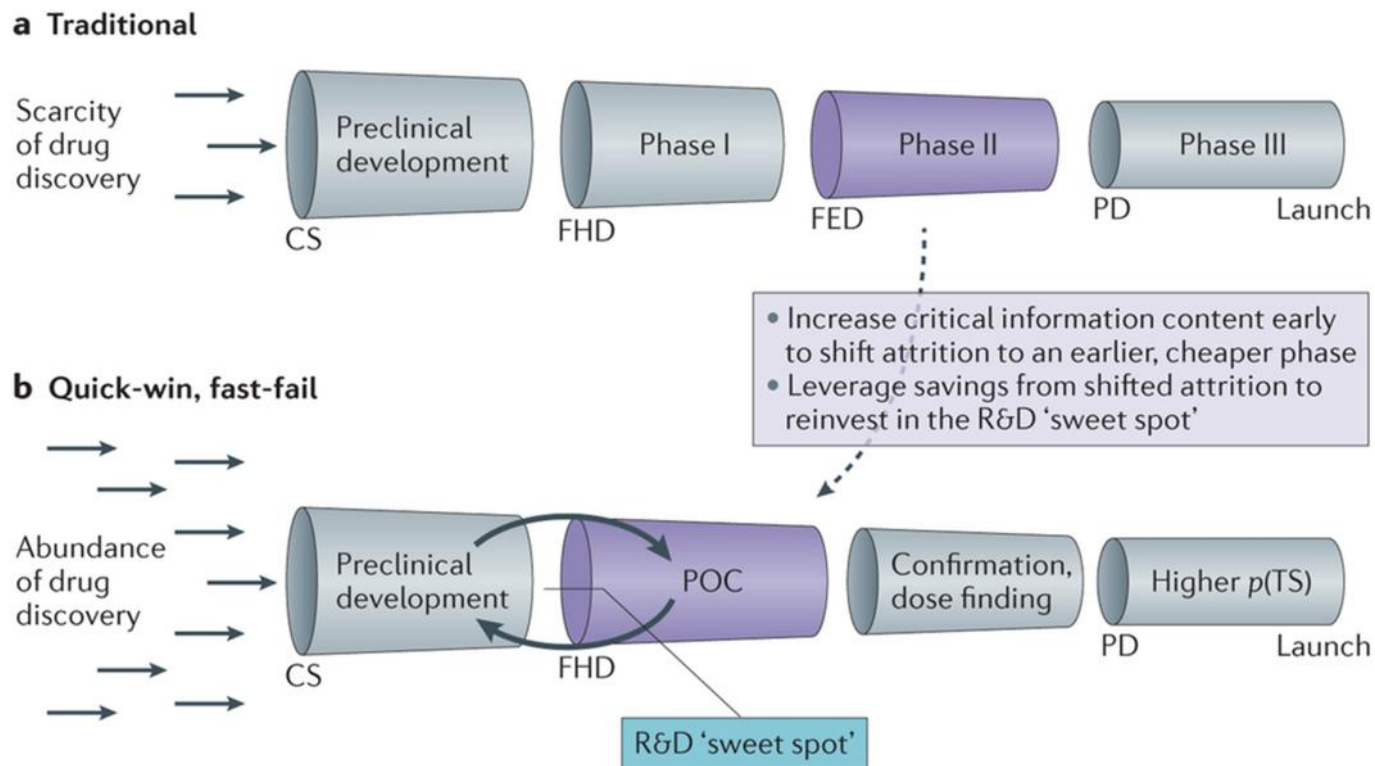
P= Productivity; WIP=Work in process; **p(TS)**=Probability of technical success; V=Value; CT=Cycle Time; C=Cost



Discovery Development

Paul et al. Nat Rev Drug Disc (2010), 9, 2003

Quick-win, fast fail



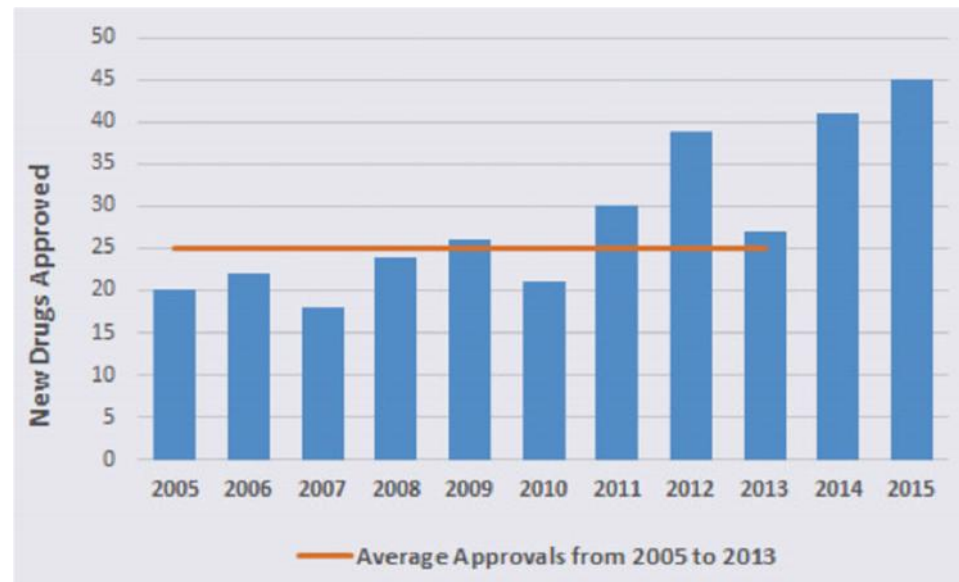
Owens et al, Nature Reviews Drug Discovery 14, 17–28 (2015)

Nature Reviews | Drug Discovery

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How many submission per year to FDA?

- Last year seem to see a new trend: finally out of “Ice Age” of pharma industry?



Maybe due to lots of first-in-class (get high chance of approval) and other FDA approved schemes **Fast Track, Breakthrough Therapy, Accelerated Approval, Priority Review**
(<http://www.fda.gov/forpatients/approvals/fast/ucm20041766.htm>)

<http://www.impactpharma.com/blog/record-numbers-of-fda-approved-drugs/>

Playing tricks

- Accelerating drugs through market

- (The Wall Street Journal via [NewsPoints Desk](#))

Drug Firms Buy \$67.5 Million Voucher to Speed FDA Review - (The Wall Street Journal via [NewsPoints Desk](#))



(Ref: [The Wall Street Journal](#))
July 31st, 2014

BioMarin got a voucher for contributing with a drug for a unmet medical need in paediatric area
Sanofi bought it for cholesterol reducing drug

Sanofi, Regeneron's PCSK9 inhibitor Praluent gains FDA priority review



(Ref: [Yahoo!Finance](#), [Morningstar](#), [FinanzNachrichten](#), [Forbes](#), [PR Newswire](#), [Bloomberg](#))

January 26th, 2015

By: Matthew Dennis

Tags: [Top Story](#) [evolocumab](#) [Praluent](#) [Amgen](#) [Regeneron](#) [Sanofi](#) [FDA](#) [Dyslipidemia](#)
[Regulatory Affairs](#)

Where all this got us

Cost and Time

- Cost of single drug is estimated to be around 1Billion \$!
- Time of getting a new drug is 13.5 years
- Lots of failures

Data generated

- Compound libraries of millions of compounds, characterized, stocked, tested
- Combinatorial chemistry
- High throughput screening facilities
- Large databases, from chemical structure, to storage, to batchID, to experiment, to 3D structures, ADME/tox

Are we using data in the right way?

- In-house data: are we looking to the data we already have in the right way?
- External data: are we accessing all the data which sits outside (institutions, companies) ?
- Data analytics offers now great opportunities: is it the case to teach a old dog (pharma) a new trick (data science)?

Digging in-house data






Drug Discovery Today

Volume 20, Issue 6, June 2015, Pages 652–658



Feature

Extending kinome coverage by analysis of kinase inhibitor broad profiling data

Edgar Jacoby¹,  , Gary Tresadern¹, Scott Bembek², Berthold Wroblowski¹, Christophe Buyck¹, Jean-Marc Neefs¹, Dmitrii Rassokhin³, Alain Poncelet⁴, Jeremy Hunt⁵, Herman van Vlijmen^{1, 6}, 

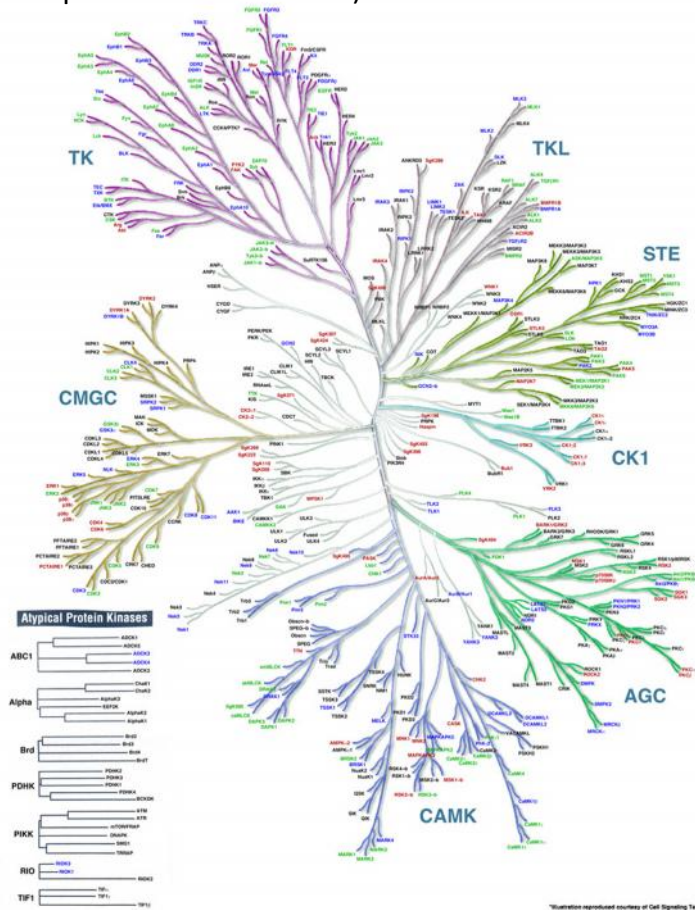
- Janssen has an extensive compound library
- Over 40 project on Kinases in the years (~1.5billions\$)
- 70K compounds synthesized to target kinases
- Can we capitalize on this gigantic effort to find new targets?

The kinome: more than 500 similar proteins

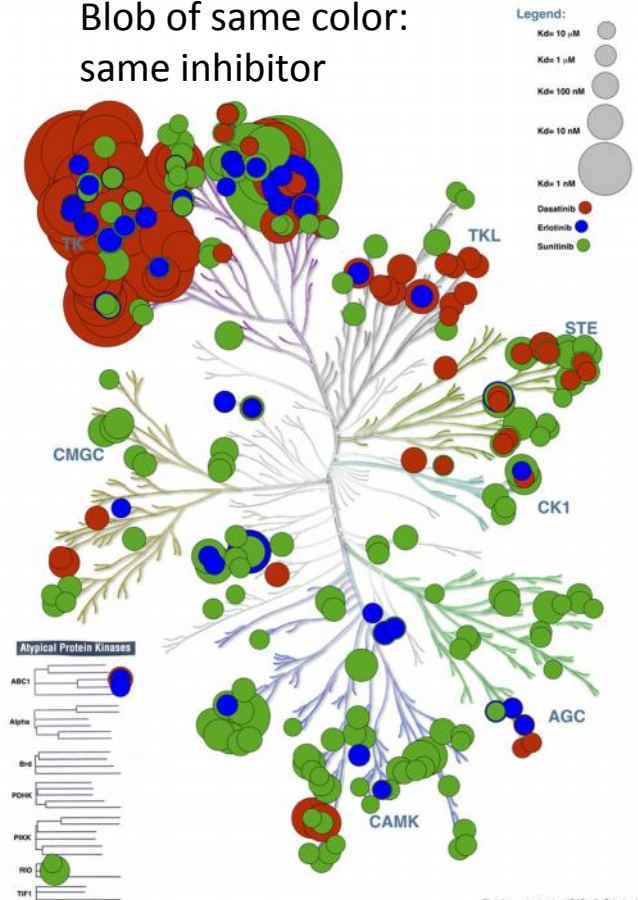
Branching is a divergence in sequence of the protein (i.e. the composition of the ribbon)

Legend:
■ Protein in PDB database
■ Protein used in Karaman et al. 2008
■ Protein in both Karaman and the PDB database

- Specificity is important to limit the side effects
- For very similar proteins a limited degree of promiscuity is inevitable
- There are also a number of well documented classes of drugs
- Mostly linked to cancer therapies
- Finding new drugs with specificity of this kind would be already a success



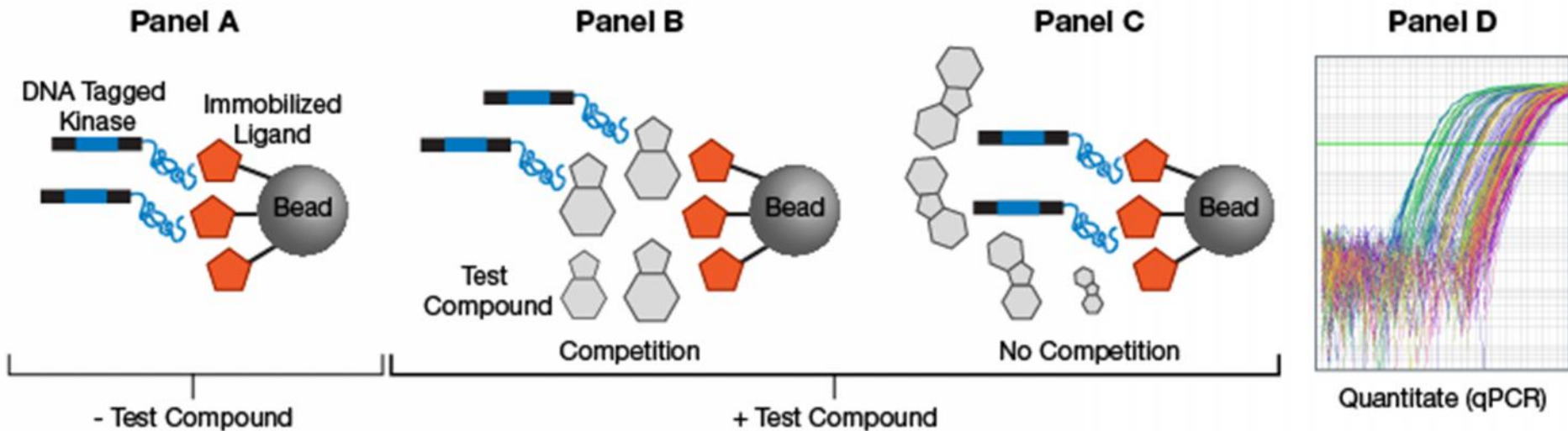
Blob of same color:
 same inhibitor



Chartier M, Chénard T, Barker J, Najmanovich R. (2013) Kinome Render: a stand-alone and web-accessible tool to annotate the human protein kinome tree. *PeerJ*:e126

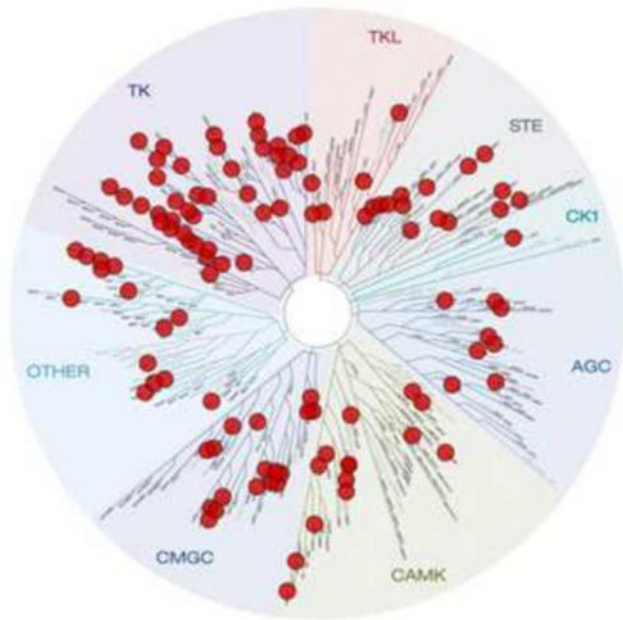
DiscoverX Kinome Scan

450 Kinases provided by DiscoverX* **3K compounds from Janssen** =13500K experiments? No!
-> Test each ligand with multiple kinases, then measure which kinases are attached on the bead. The ones which are not **attached have interacted with the ligand**



<https://www.discoverx.com/technologies-platforms/competitive-binding-technology/kinomescan-technology-platform>

Results



KD/IC50 \leq 10 nM S65 \leq 0.05

129 kinases by 434 cpds

- New potent and selective compounds for many new kinases are found (55)
- New projects were started as a consequence of this effort
- Much of these compounds is already known beforehand since they've been amply characterized (cost/time cut)
- Good eye for new technologies provide new ways to benefit from material and data already present in house

Share with care: pre-competitive agreements

According to Pistoia Alliance: “**aggregating, accessing, and sharing data that are essential to innovation, but provide little competitive advantage**”.

Companies and Institutions put some data in a third party institution which act as a broker for the projects of each contributor **to protect everyone’s intellectual property**

- Innovative Medicines Initiative
 - ETOX
 - K4DD
 - EMIF
 - OpenPhacts
 - European Lead Factory
 - Etc....
- MedChemica SALT



European Lead Factory



- Idea: my competitor has compounds is not interested anymore. May I speed up my research by using them? After all, he is not much interested anymore in them!
- 30 Institutions and companies share proprietary compounds
- 500K compounds ready to be screened
- Facilities in Scotland (compound library) and the Netherlands for screening
- Scientists who contribute with novel compounds are rewarded
- Researchers and companies can ask to test a target against the compound collection
- Only the confirmed active (~50) will be shared and all the rest of the screened compounds remain unknown to protect the IP of those who shared the compounds
- Companies share lots of knowledge but they disclose very little at a time while having huge impact on productivity

Digging in others' data: loads of info in the outside world

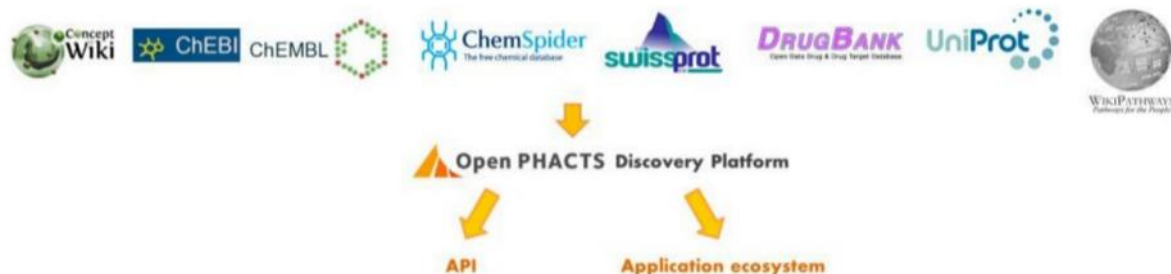
- ChEMBL: 1,686,695 cpds, annotated
- PubChem: 82 millions cpds
- SwissProt: protein database
- Uniprot: protein databse
- Genebank
- Literature



OpenPhacts



- Lots of databases: drugs, properties, proteins, genes
- Their information is somewhat connected but each single Pharma does the effort to integrate it: redundant efforts
- Task: create a “semantic integration hub”, a common standard and API where company can access all those data and integrate their own
- Integrate informations about compound–target–pathway–disease/phenotype
- 31 academics, 9 pharma industries, 3 software SME



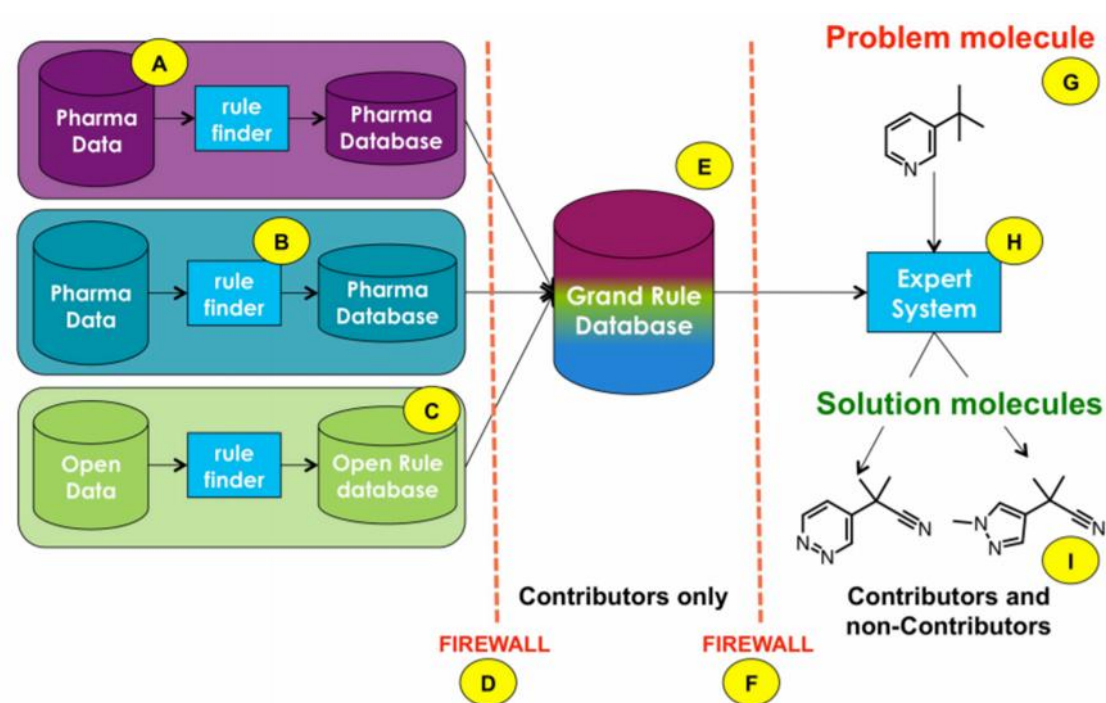
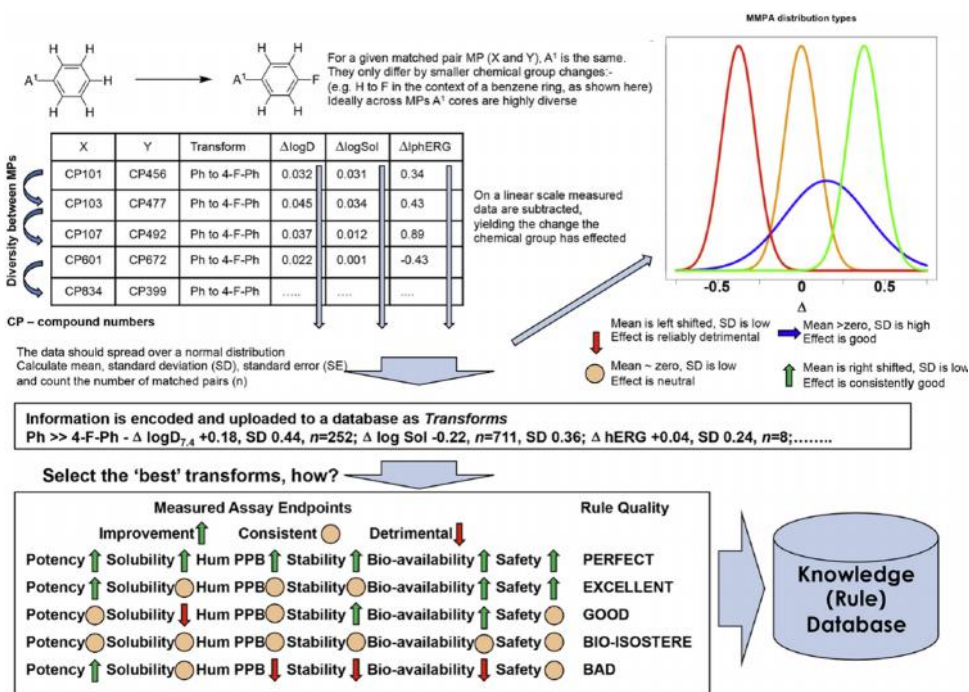
Review

Scientific competency questions as the basis for semantically enriched open pharmacological space development

Kamal Azzaoui¹, Edgar Jacoby¹⁴, Stefan Senger², Emiliano Cuadrado Rodriguez³, Mabel Loza³, Barbara Zdrzil⁴, Marta Pinto⁴, Antony J. Williams⁵, Victor de la Torre⁶, Jordi Mestres⁷, Manuel Pastor⁷, Olivier Taboureau⁸, Matthias Rarey⁹, Christine Chichester¹⁰, Steve Pettifer¹¹, Niklas Blomberg^{12, a}, Lee Harland¹³, Bryn Williams-Jones¹³, Gerhard F. Ecker¹  

MedChemica SALT: a privately owned

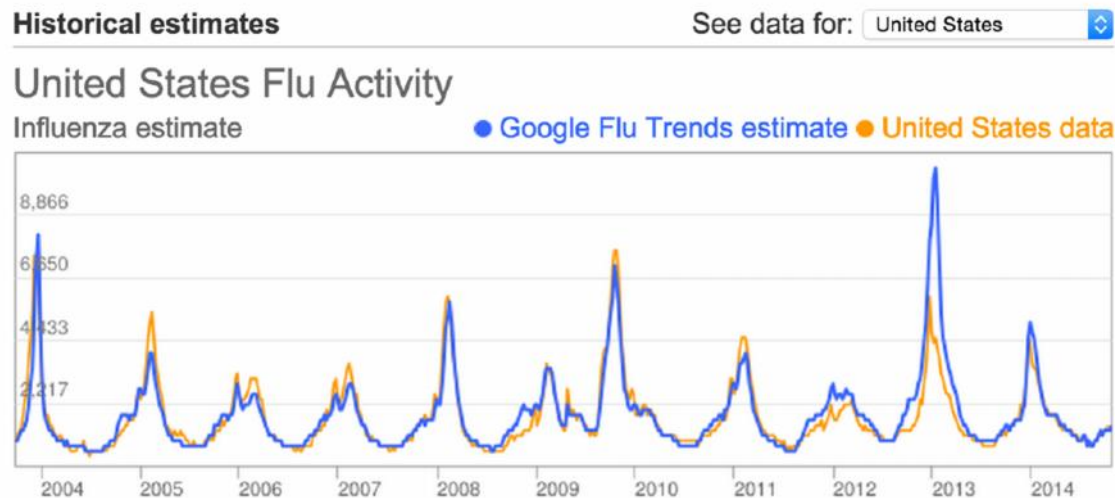
- Based on Matched Molecular Pairs



Dossetter et al Drug Discovery Today (2013) 18, 724–731

Google FLU Trends

- Launched in 2008 now closed
- Based on google user searches for terms related to flu
- Could predict the spread of flu few weeks ahead respect to the Center for Disease Control and Prevention
- This can help adjusting medical support logistics



United States: Influenza-like illness (ILI) data provided publicly by the [U.S. Centers for Disease Control](http://www.cdc.gov).

“Correlation supersedes causation, and science can advance even without coherent models, unified theories, or really any mechanistic explanation at all.”

Chris Anderson, Wired, 2008

<https://www.wired.com/2008/06/pb-theory/>

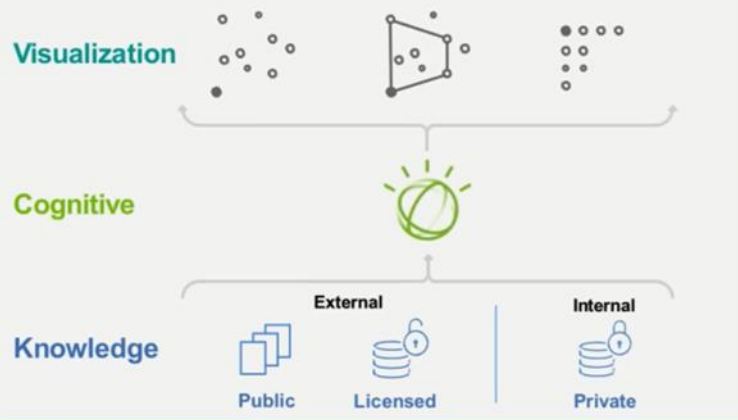
And of course IBM Watson...

Watson for Drug Discovery

Watson for Drug Discovery: Accelerating Discovery

Watson for Drug Discovery is a **cloud-based, end-to-end scalable platform** that helps life science researchers potentially discover new disease pathways, new drug targets and additional drug indications

Three Core WDD Components



The Watson Advantage

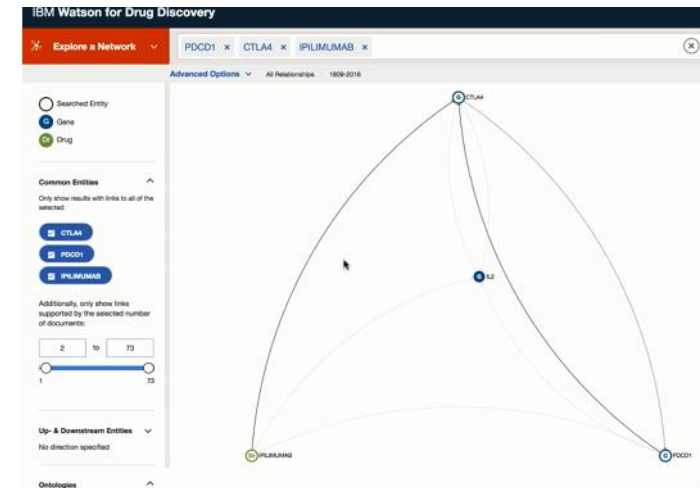
Key Capabilities

- **Aggregates** diverse content
- **Cognitive** technology
- **Scalability**
- **Domain expertise**
- **Agility / speed**

Key Benefits

Helps:

- **Accelerate** insight g
- **Improve** researcher **productivity**



Canadian neuroscience leaders tap IBM Watson to speed time to discovering new drugs for Parkinson's disease

IBM Watson for Drug Discovery chosen to help researchers more rapidly pinpoint promising drug targets

Uses Natural Language Processing to dig out quickly unsuspected relations occurring in literature to highlight new targets, biomarkers etc.

Info Hub

News releases

Info Hub archives

News release archives

TORONTO, ON – 12 October, 2016: IBM (NYSE: IBM) Watson Health today announced that the Ontario Brain Institute (OBI) and the Movement Disorders Clinic (MDC) at Toronto's University Health Network (UHN) will embark on Canada's first ever Parkinson's disease research project using the recently launched IBM Watson for Drug Discovery.

Genomics revolution

- Deciphering the human genome took 10 years and 3 Billion \$
- Now a whole genome can be mapped in 24 hours and costs around 1000\$

nature

Vol 437|15 September 2005|doi:10.1038/nature03959

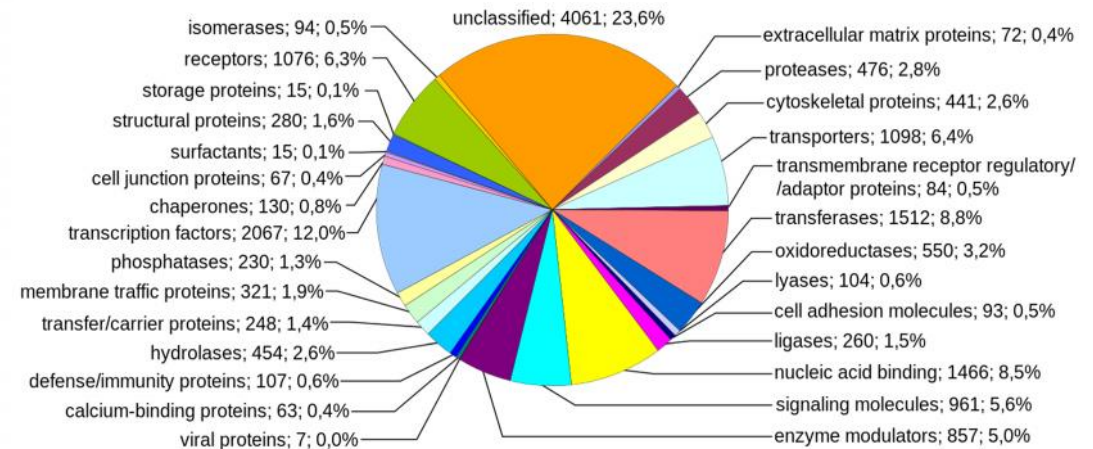
ARTICLES

Genome sequencing in microfabricated high-density picolitre reactors

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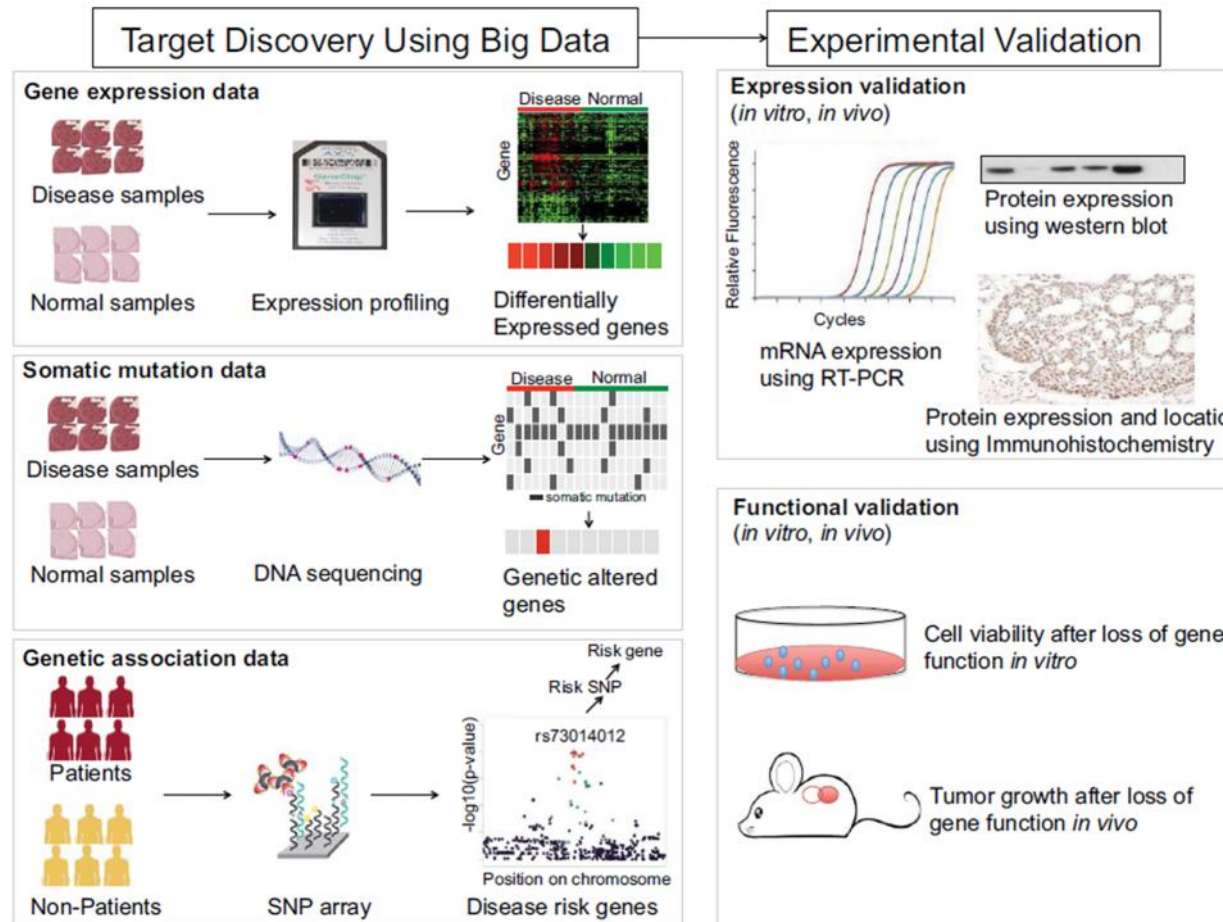
The proliferation of large-scale DNA-sequencing projects in recent years has driven a search for alternative methods to reduce time and cost. Here we describe a scalable, highly parallel sequencing system with raw throughput significantly greater than that of state-of-the-art capillary electrophoresis instruments. The apparatus uses a novel fibre-optic slide of individual wells and is able to sequence 25 million bases, at 99% or better accuracy, in one four-hour run. To achieve an approximately 100-fold increase in throughput over current Sanger sequencing technology, we have developed an emulsion method for DNA amplification and an instrument for sequencing by synthesis using a pyrosequencing protocol optimized for solid support and picolitre-scale volumes. Here we show the utility, throughput, accuracy and robustness of this system by shotgun sequencing and *de novo* assembly of the *Mycoplasma genitalium* genome with 96% coverage at 99.96% accuracy in one run of the machine.

Human genome is 3 Billions of base pairs!



Publicly available genome data: finding new targets

Detect a change in a known pattern (i.e. overexpression, mutation), consider high noise is expected (for full genome sequence) and druggability (i.e. has pockets for small mols in relevant regions) of the target must be considered



Validate the causal relation



May found that more than one drug is needed



Drug repurposing + personalized medicine!

Conclusions

- Technology innovation and integration is key in pharma industry
- Many branches of today's most exciting science under the same hood: chemistry, (molecular)biology, genomics, physical chemistry, molecular modeling
- Data integration plays a central role
- Standardization is ongoing
- New technologies appear, these give new opportunities to use existing data
- Using external data is becoming appealing too, in particular when using third parties that protect knowhow