



A novel platform to identify synergistic drug combination to combat cancer recurrence and prolong clinical efficacy

ChemAxon User Meeting, Cambridge MA, March 25-26 2019

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Acknowledgements



Nagi Ayad

- Amar Koleti
- Dušica Vidović
- John Turner
- Tanya Kelley
- Afoma Umeano
- Derek Essegian
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- Caty Chung
- Daniel Cooper
- Rimpi Khurana

- Avi Ma'ayan
- Mario Medvedovic
- Stephan Schürer

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BD2K-LINCS DCIC U54 HL127624





BY THE NUMBERS

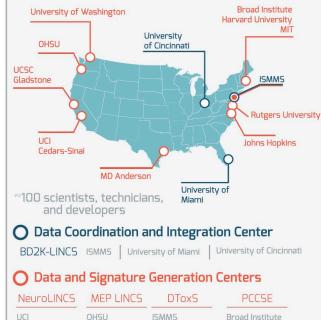
LINCS

Cedars-Sinai

Johns Hopkins

Glastone

15 INSTITUTIONS



Harvard University

Rutgers University University of Washington MD Anderson

HMS LINCS **Broad Transcriptomics**

Broad Institute

45 Primary Cells

Differentiated Cells

5 CELL TYPES

31 Cell Lines

4 PERTURBATION TYPES

18 IPSCs

41.847

7,547

Genetic

Perturbations

Small Molecules

2 Embryonic Stem Cells

2.736

Microenvironment

Perturbagens

Imaging 11 Assays **85** Datasets

HMS LINCS Database (A)

Slicr

ILINCS

piLINCS

Harmonizome

Download LINCS Datasets

0

Transcriptomics

Binding

2 Assays

11 Datasets

2 Assays

701 Datasets

4 Assays

5 SIGNATURE TYPES



















Proteomics











LINCS Data Portal

OmicsIntegrator



L1000CDS



Enrichr SEP - L1000

Breast Cancer Browser

30 MORE TOOLS

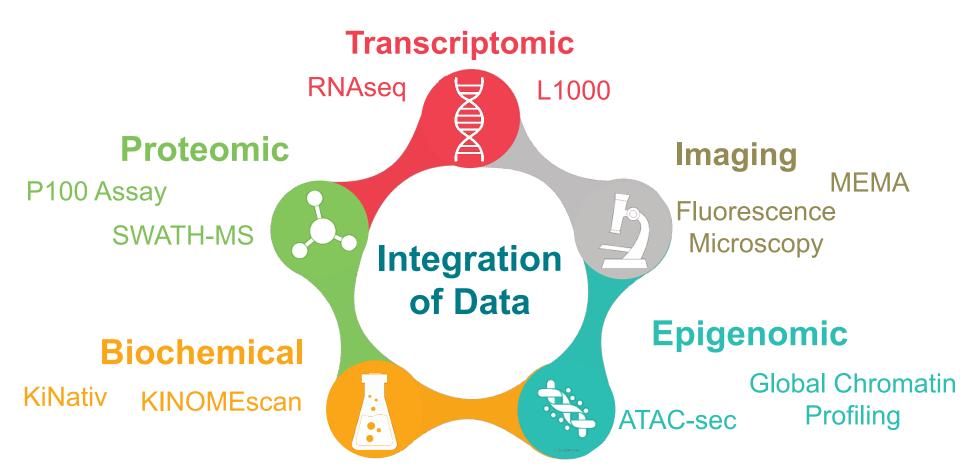




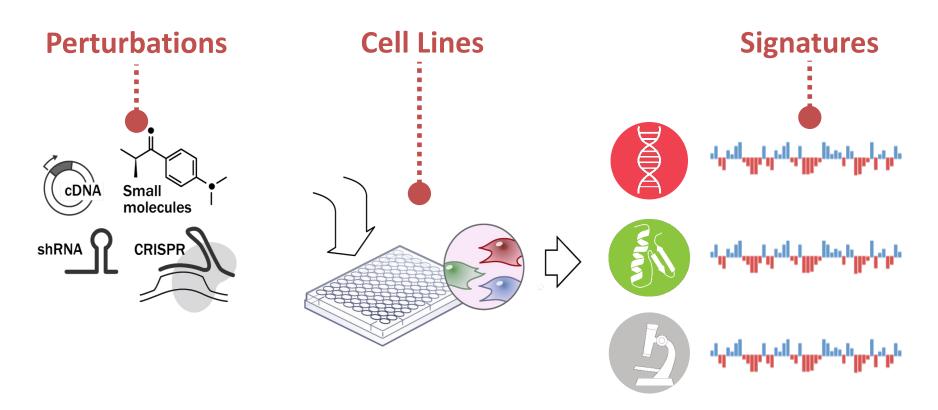




LINCS Project generates diverse multidimensional data



LINCS Signatures characterize diverse cellular responses upon perturbation



LINCS Data Portal - Home Page













381 Datasets

41847 Small Molecules

В

1127 Cells

978 Genes

1469 Proteins /155 Peptide Probes 8 Antibodies

2 Datasets

14 Methods

06 Subject Areas

12 Centers

06 Projects

11 Biological Processes

inding	201	Datasets

Imaging 85 Datasets

Transcriptomics 12 Datasets

Proteomics 9 Datasets

Epigenomics 4 Datasets

Metabolomics 1 Datasets

KINOMEscan	kinase-small	molecule	binding assay
KIIIOIVILSCAII	VIIIaac-2IIIaii	IIIOIECUIE	Dillully assay

ELISA protein state assay

	163 Datasets
KiNativ kinase-small molecule binding assay	30 Datasets
Bead-based immunoassay for protein state	3 Datasets
ELISA protein secretion profiling assay	3 Datasets

KINOMEscan kinase-small molecule binding assay

Assay Overview

Compounds that bind the kinase active site and directly (sterically) or indirectly (allosterically) prevent kinase binding to the immobilized ligand, will reduce the amount of kinase captured on the solid support (Panels A & B). Conversely test molecules that do not bind the kinase have no effect on the amount of kinase captured on the solid support (Panel C). Screening hits are identified by

LINCS Data Portal - http://lincsportal.ccs.miami.edu/



Small Molecules

Accelerate the discovery of novel therapeutics for precision medicine by unravelling LINCS small molecules annotations for approved drugs, clinical phase and chemical probes.

Search assays, cell lines, and perturbagens..

Search for Small Molecule

Search LINCS small molecules by generic names, synonyms, trade names and explore across LINCS datasets, model systems and annotations.

Example search for Seliciclib can query using either "Roscovitine" or "CYC-202". A comprehensive list of generic names and synonyms are also stored and can be used as search terms.

Explore Annotations

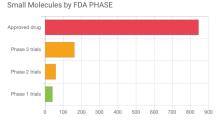
Search can be performed by a wide set of annotations such as mechanism of actions Cyclin-dependent kinase 2 inhibitor, MESH terms CARCINOMA, EFO terms CYSTIC FIBROSIS and cells MCF7.

Advanced Search

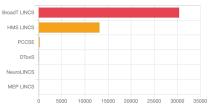
Search also facilitates AND, OR, NOT queries. Some examples 1) find Small Molecules that are used in KINOMEscan and L1000 assays. 2) Find Small Molecules that have annotated as Cyclin-dependent inhibitor or kinase inhibitor.

Browse | Structure Search | About | Help

Statistics









Cells

Accelerate the discovery of novel therapeutics for precision medicine by unravelling LINCS cells annotations.

Search assays, cell lines, and perturbagens...

Browse | About | Help

Search for Cells

Search LINCS cells by generic names, synonyms, diseases and explore across LINCS datasets, perturbagens and annotations.

Example search for MCF7 or explore more details through the Cell Landing Pages "MCF7". A comprehensive list of generic names and synonyms are also stored and can be used as search terms.

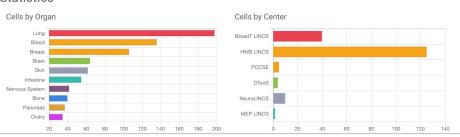
Explore Annotations

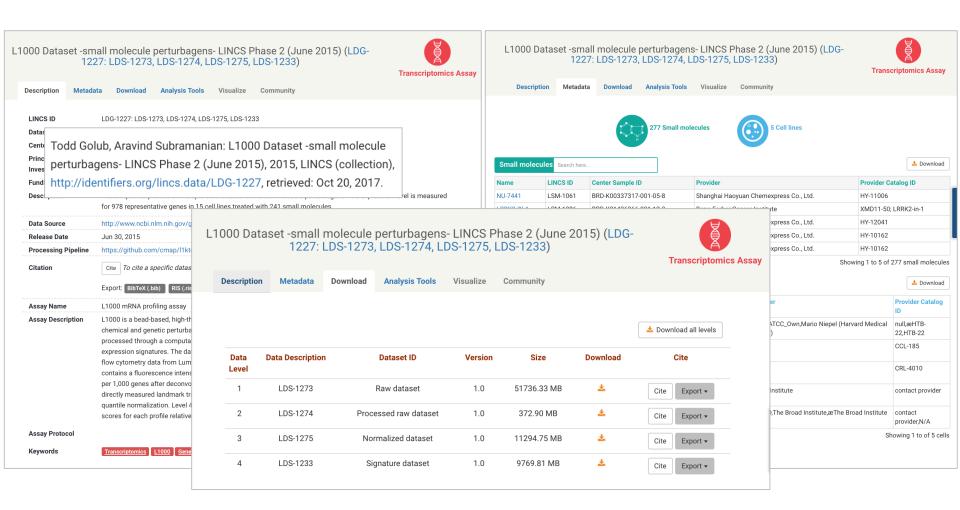
Search can be performed by a wide set of annotations such as organs Breast, Diseases Melanoma, Assays Fluorescence imaging cell growth inhibition assay and Small Molecules Enzastaurin.

Advanced Search

Search also facilitates AND, OR, NOT queries. Some examples 1) find cells that are used in P100 and L1000 assays. 2) Find cells that are derived from Lung.

Statistics









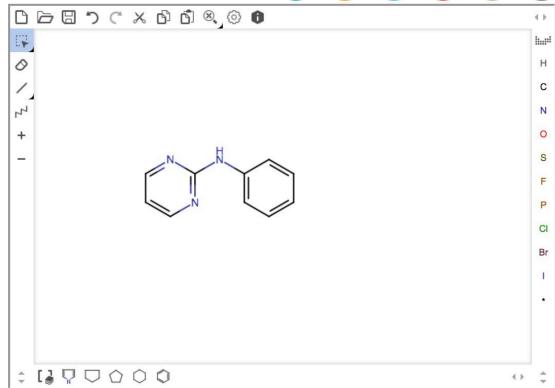








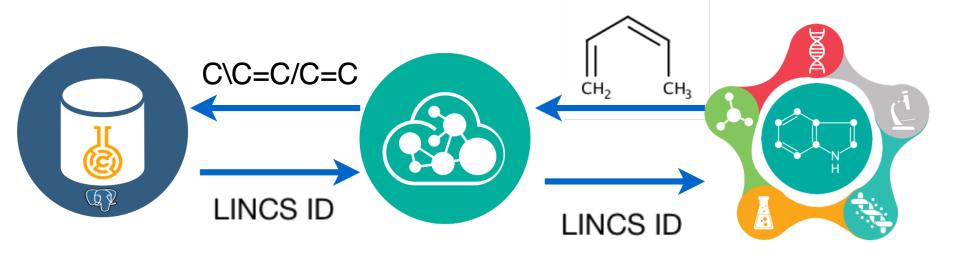




Substructure Search Similarity Search

POWERED BY ChemAxon

LDP Chemical Structure Search Engine



SMDB

JChem Postgres
Cartridge

API Servlet

Structure
Search
Marvin JS or direct

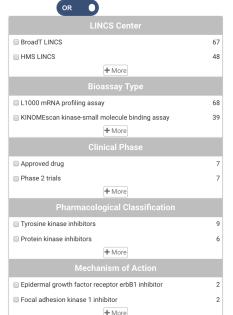












Search assays, cell lines, and perturbagens. Filtered by: LSM-15449" OR "LSM-21027" OR "LSM-24505" OR "LSM-1459" OR "LSM-18669" OR "LSM-4750" OR "LSM-36387" O (lincsidentifier) × Total Small Molecules: 93 ■ List view **Imatinib** Transcriptomics Synonyms: CGP-57148B; Gleevec; Glivec; Imatinib; IMATINIB (3524741); Imatinib Mesilate; Imatinib Mesylate; Binding IMATINIB (STI571); STI-571 Metabolomics Max Phase: Approved drug Proteomics Mechanism of Action: Platelet-derived growth factor receptor beta inhibitor; Stem cell growth factor receptor Imaging inhibitor; Tyrosine-protein kinase ABL inhibitor Epigenomics Pharmacological Classification: Tyrosine kinase inhibitors; Protein kinase inhibitors Model Systems: 669 cell-lines used Show ✓ Explore 🗗 Datasets: Used in 32 Datasets Show V **Nilotinib** Transcriptomics Synonyms: AMN-107; Nilotinib; NILOTINIB (AMN-107); Nilotinib Hydrochloride Monohydrate; Tasigna Binding Max Phase: Approved drug Metabolomics Mechanism of Action: Tyrosine-protein kinase ABL inhibitor Proteomics Pharmacological Classification: Tyrosine kinase inhibitors; Protein kinase inhibitors Imaging Model Systems: 162 cell-lines used Show ▼ Explore ☑ Epigenomics Datasets: Used in 31 Datasets Show V 15 **TAE-684** Transcriptomics Synonyms: NVP-TAE684 Binding Model Systems: 637 cell-lines used Show ✓ Metabolomics Datasets: Used in 28 Datasets Show V Proteomics Imaging Epigenomics 10 20



BI-2536

Synonyms: BI 2536; BI-2536; NPK33-1-98-1

Max Phase: Phase 2 trials

Mechanism of Action: Serine/threonine-protein kinase PLK1 inhibitor
Explore ☑ Model Systems: 644 cell-lines used Show ✓ Explore ☑ Datasets: Used in 27 Datasets Show V

Transcriptomics

Binding Metabolomics Proteomics Imaging



Transcriptomics

Metabolomics

Proteomics

Epigenomics

Imaging

Binding





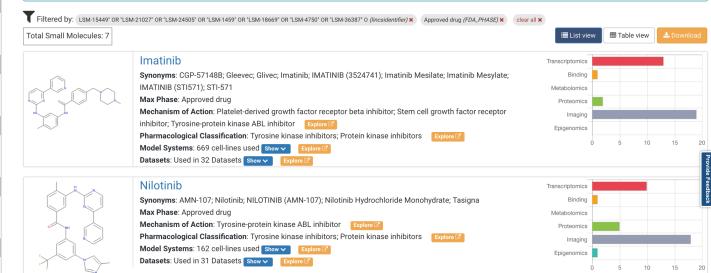


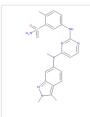




+ More

Search assays, cell lines, and perturbagens...





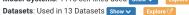
Pazopanib

Synonyms: 635702-64-6; GW786034; GW786034B; Pazopanib; Pazopanib HCl; Pazopanib Hydrochloride; Votrient Max Phase: Approved drug

Mechanism of Action: Fibroblast growth factor receptor 3 inhibitor; Macrophage colony stimulating factor receptor inhibitor; Tyrosine-protein kinase LCK inhibitor; Tyrosine-protein kinase ITK/TSK inhibitor; Vascular endothelial growth factor receptor inhibitor; Stem cell growth factor receptor inhibitor; Platelet-derived growth factor receptor inhibitor; Fibroblast growth factor receptor 1 inhibitor

Pharmacological Classification: Angiogenesis inhibitors; Protein kinase inhibitors

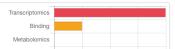
Model Systems: 1115 cell-lines used Show ▼ Explore ☑





Osimertinib

Synonyms: AZD-9291; AZD-9291 MESYLATE; HG-14-8-02; Osimertinib; Osimertinib Mesylate; Tagrisso Max Phase: Approved drug















Imatinib (LSM-1023)

Description

Datasets

Curated Data

Compound Information

Name	Imatinib
LINCS (ID)	LSM-1023
PubChem (CID)	5291
ChEBI (ID)	31690 , 45783
ChEMBL (ID)	CHEMBL1642, CHEMBL2386595, CHEMBL941
Alternative Names	CGP-57148B; Gleevec; Glivec; Imatinib; IMATINIB (3524741); Imatinib Mesilate; Imatinib Mesylate; IMATINIB (STI571); STI-571
SMILES 10	Show ✓
InChI	Show

Find Similar Structures | Similarity Search -

POWERED BY & ChemAxon

95% Similar Molecules

90% Similar Molecules

80% Similar Molecules

Related LINCS Compounds by chemical structure (similarity ≥ 0.9)

LSM-43288			
LSM-45329			
LSM-44525			

Predicted Physicochemical Properties

Molecular formula	C29H31N7O
Molecular weight	493.60g/mol
Lipinski 3	Fail
Lipinski 5	Pass
Bioavailability	Pass
Lead likeliness	Fail

cLogP	4.38
Rotatable bonds	7
H bond acceptors	7
H bond donors	2
TPSA	86.28
QED	0.40

LINCS Data Portal | Small Molecules



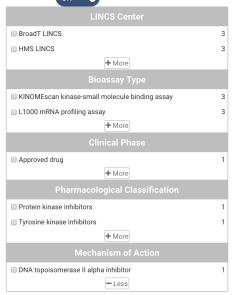




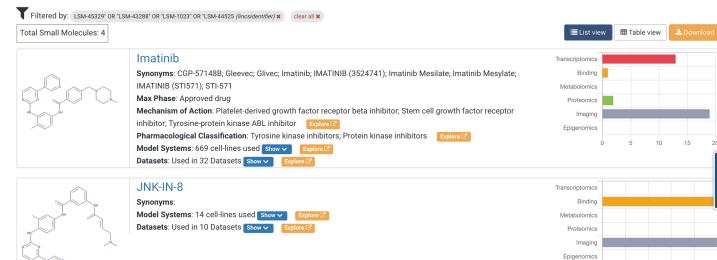


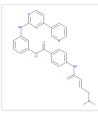






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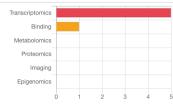




JNK-IN-2

Synonyms: ZG-10

Model Systems: 93 cell-lines used Show > Datasets: Used in 6 Datasets Show V



Glioblastoma Multiforme (GBM)



- Most common and aggressive adult brain tumor
- Occurs primarily in ages of 45-70 years



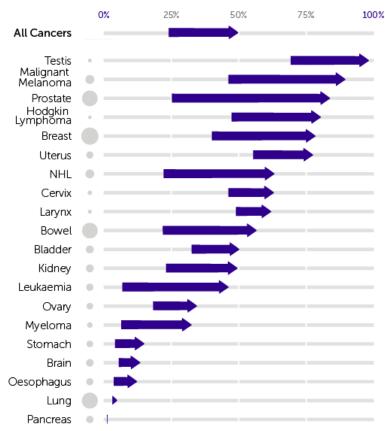
 Surgical resection followed by radiotherapy and temozolamide chemotherapy



- Annual incidence:1 / 19,000
- 5-years survival rate< 9.8%

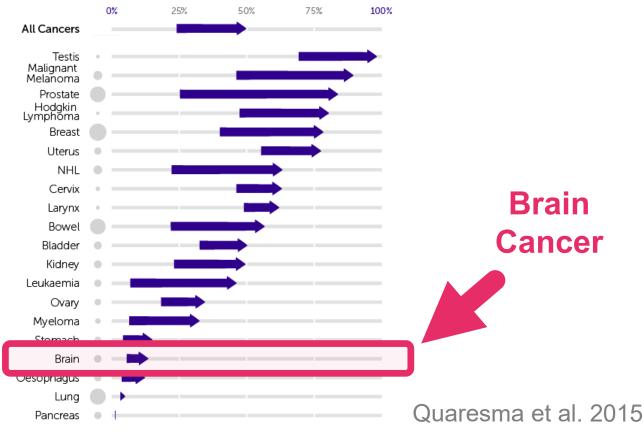
Little progress has been made in the survival of brain cancer patients



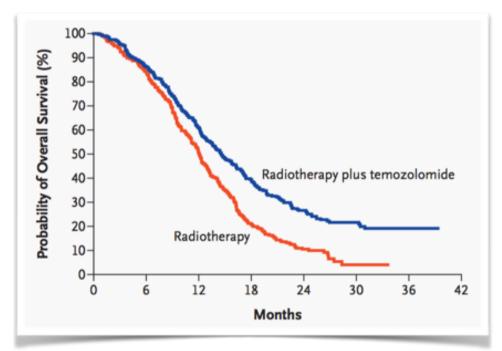


Little progress has been made in the survival of brain cancer patients





Current GBM Treatment is Ineffective



Standard of Care

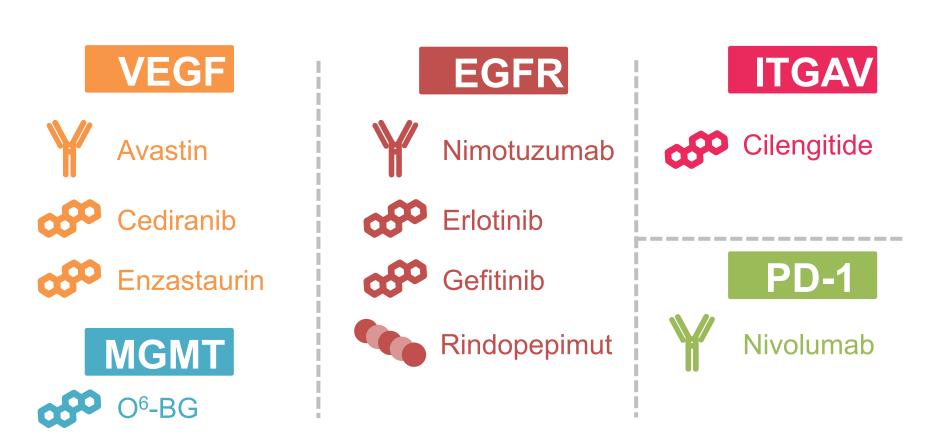
- Maximal surgical resection
- 2. Radiotherapy
- 3. Temozolomide (TMZ) treatment

No early detection Biomarker

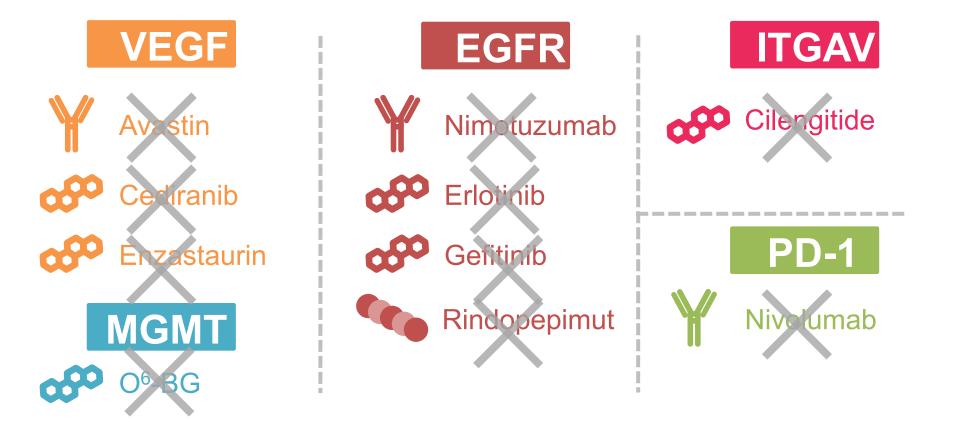
Temozolomide (2005)

2.5 month increase in survival (14.6 vs 12.1 months)

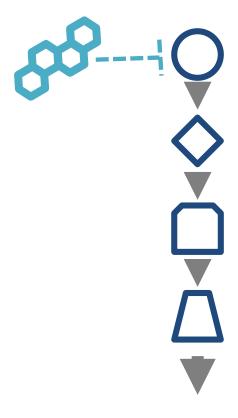
Targeted phase III trials have failed to improve overall survival of Glioblastoma patients

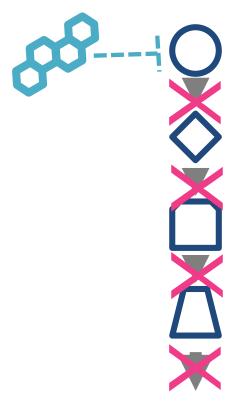


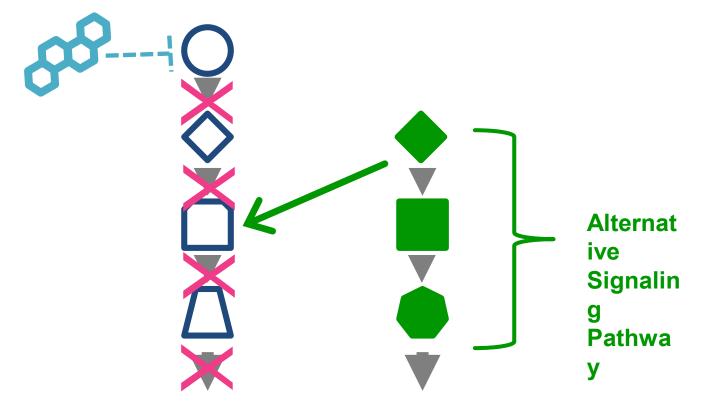
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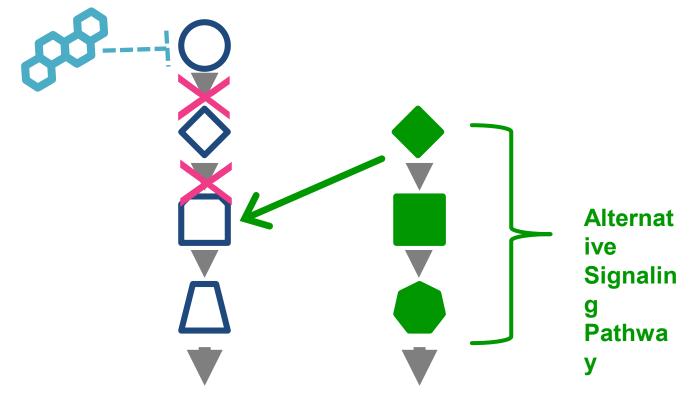


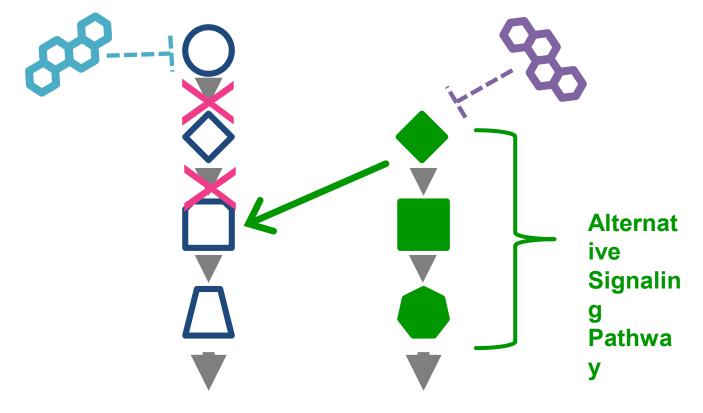


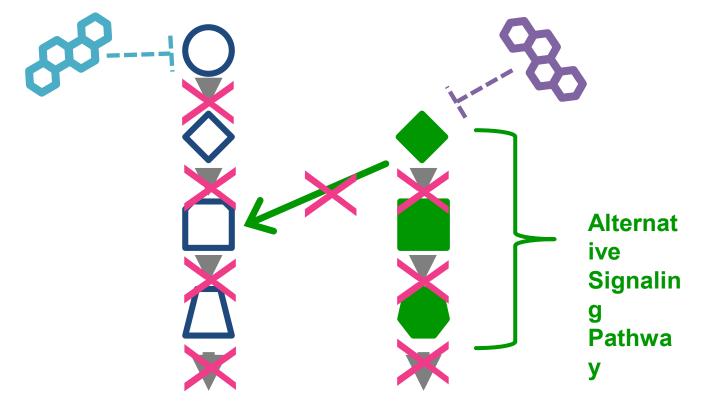




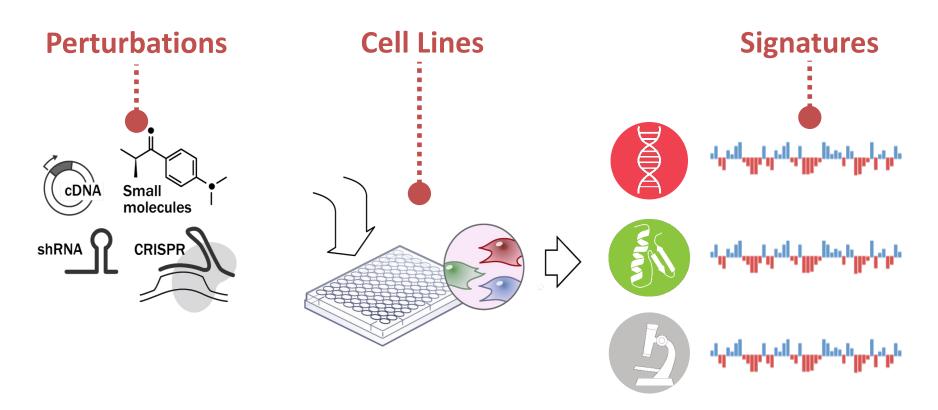




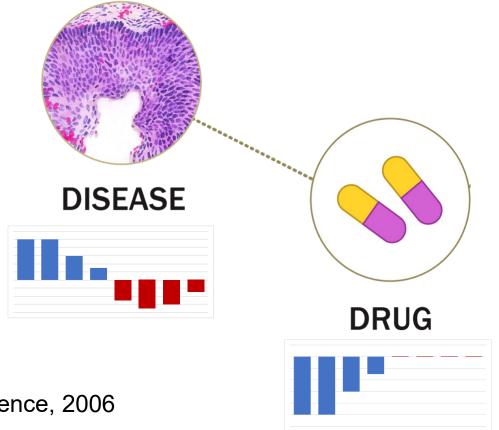




LINCS Signatures characterize diverse cellular responses upon perturbation



CMap concept: linking diseases to therapeutics



Lamb et al. Science, 2006



Received 7 Dec 2016 | Accepted 17 May 2017 | Published 12 Jul 2017

DOI: 10.1038/ncomms16022

OPEN

Reversal of cancer gene expression correlates with drug efficacy and reveals therapeutic targets

Bin Chen^{1,*}, Li Ma^{2,*}, Hyojung Paik^{1,3}, Marina Sirota¹, Wei Wei², Mei-Sze Chua², Samuel So² & Atul J. Butte¹





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SCIENCE TRANSLATIONAL MEDICINE | RESEARCH ARTICLE

CANCER

Systems biology-based drug repositioning identifies digoxin as a potential therapy for groups 3 and 4 medulloblastoma

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SCIENTIFIC REPORTS

Personalised drug repositioning for Clear Cell Renal Cell Carcinoma using gene expression

Karel K. M. Koudijs¹, Anton G. T. Terwisscha van Scheltinga¹, Stefan Böhringer², Kirsten J. M. Schimmel¹ & Henk-Jan Guchelaar¹



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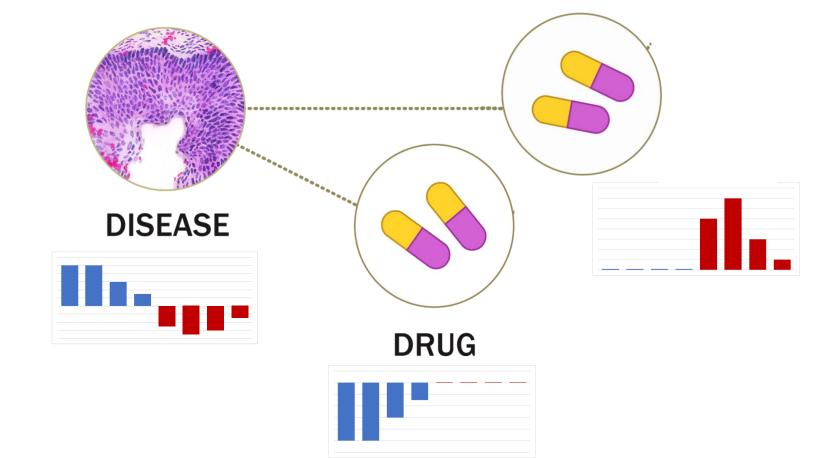
Integrated Systems and Technologies

Cancer Research

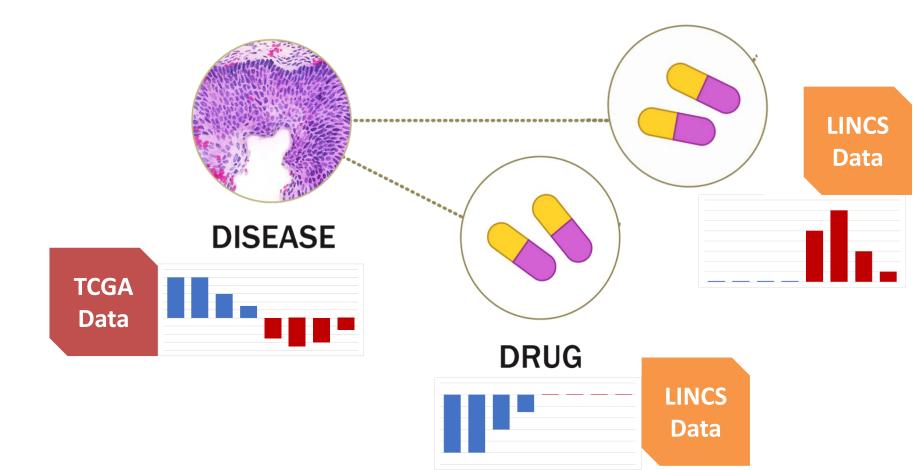
Novel Drug Candidates for the Treatment of Metastatic Colorectal Cancer through Global Inverse Gene-Expression Profiling ☑

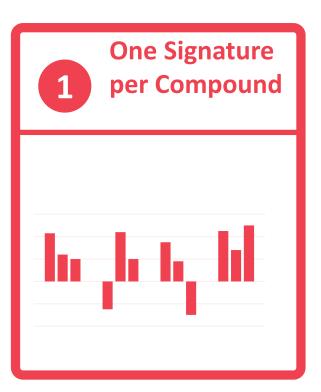
Vera van Noort^{1,2}, Sebastian Schölch³, Murat Iskar¹, Georg Zeller¹, Kristina Ostertag⁴, Christine Schweitzer³, Kristin Werner³, Jürgen Weitz³, Moritz Koch³, and Peer Bork^{1,5}

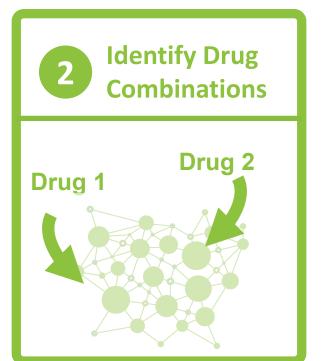
Similar approach for combination therapies



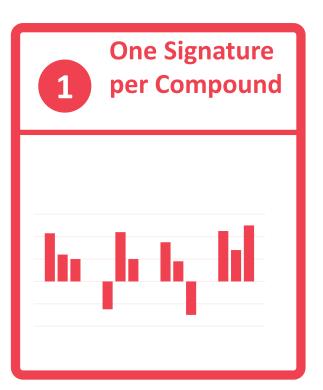
Similar approach for combination therapies

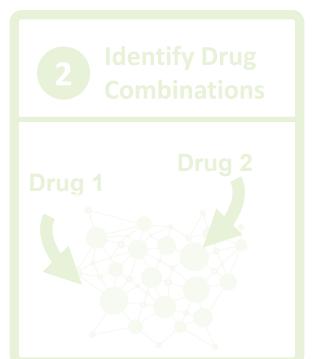






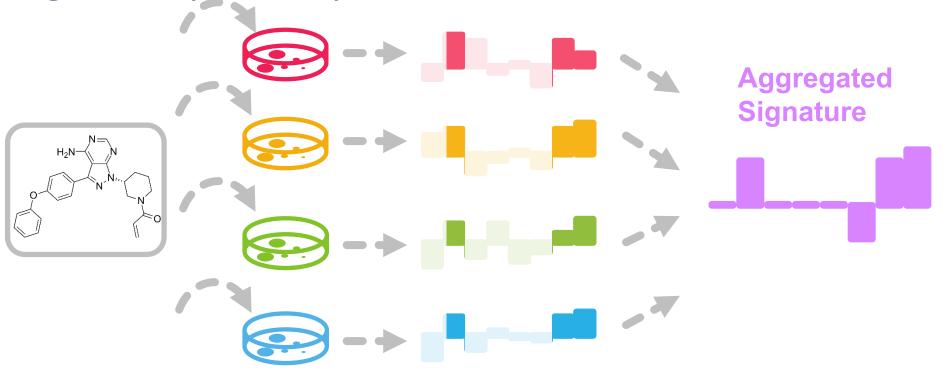




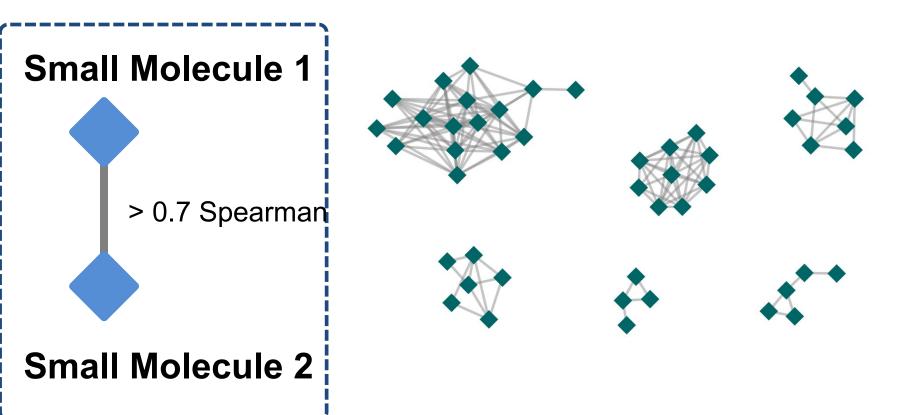




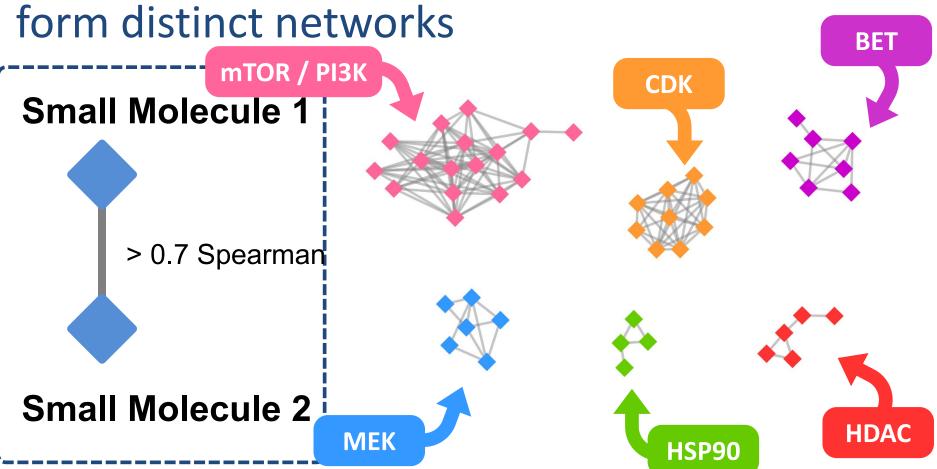
Collapse the gene expression profiles into one signature per compound



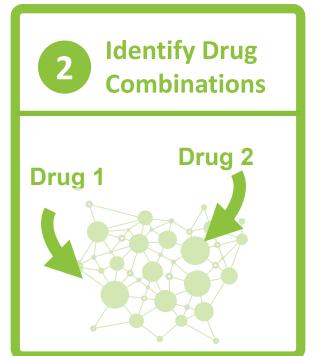
Compounds with same Mechanism of Action form distinct networks



Compounds with same Mechanism of Action

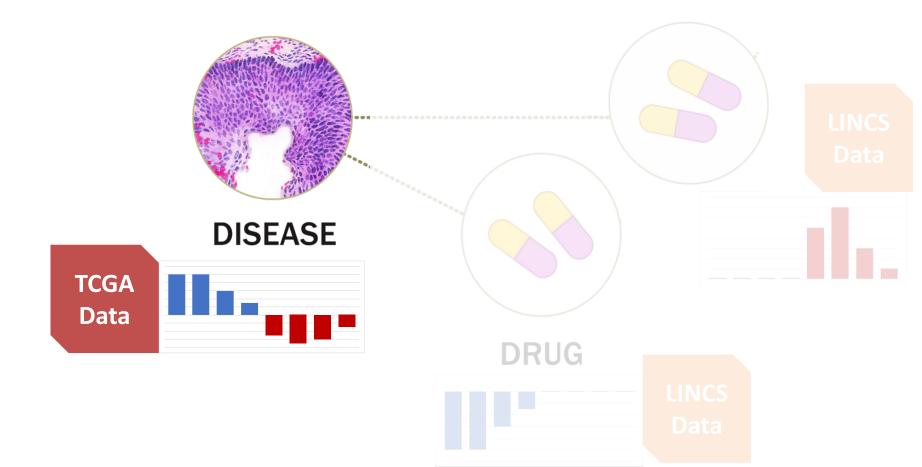




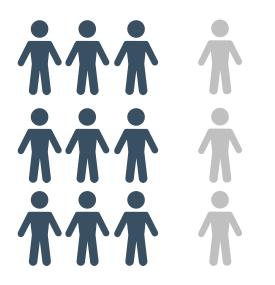




Glioblastoma disease signature from TCGA

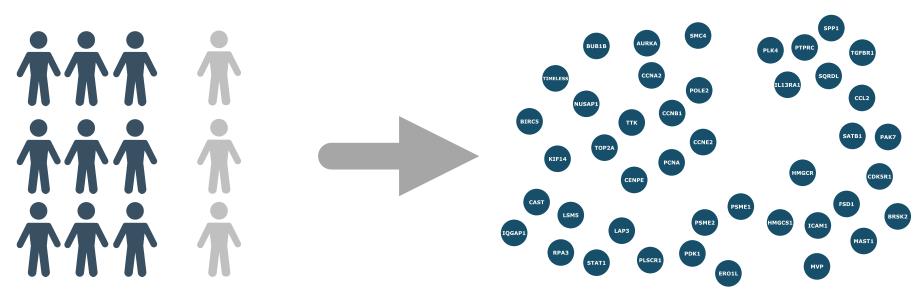


Identify Differentially Expressed Genes in Glioblastoma Samples



Glioblastoma Samples and Controls (from TCGA)

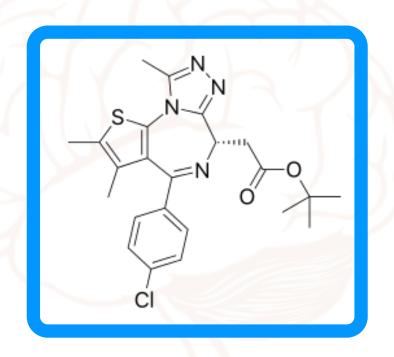
Identify Differentially Expressed Genes in Glioblastoma Samples



Glioblastoma Samples and Controls (from TCGA) Differentially Expressed Genes in Glioblastoma

JQ1 is a potent Bromodomain inhibitor

- JQ1 reduces proliferation and survival of glioblastoma PDXs
- <u>Bromodomains</u>: Readers of lysine acetylation
- Many cancers are sensitive to bromodomain inhibitors



Resistance after Bromodomain inhibition emerges in multiple cancer types

Breast Cancer

Response and resistance to BET bromodomain inhibitors in triple-negative breast cancer

Shaokun Shu^{1,2}*, Charles Y. Lin^{1,2}*, Housheng Hansen He^{1,2,3,4,5}*, Robert M. Witwicki^{1,2}*, Doris P. Tabassum¹, Justin M. Roberts¹, Michalina Jantszewska²¹, Sung Jin Huh¹², Yi Liang¹, Jeremy Ryan¹², Ernest Doherty³. Hisham Mohammed', Hao Guo², Daniel G. Stover¹², Muhammad B. Ekram¹², Guillermo Peluffo¹², Jonathan Brown³², Chris Per Dohert Brown³², Chris Brown³², Chris Brown³², Chris Han Brown³², Chris Albarda, Jun Qil², Min Ni², Prakash K. Rao², Melissa Duarte², Shwu- Yuan Wu², Cheng-Ming Chiang¹⁰, Lars Anders³, Richard A, Youngʻi. Frie P. Winera', Antony Letta¹², William T. Barry²³, Jason S. Carrolf', Henry W. Long³³, Myles Brown¹³, X. Shirley Liu³, Ni², Cifford A. Meyer³, James E. Bradner¹, Ni², Skrnelia Polyski Assia², Sao Radner², Ni², Ni², Parkash K. Rao², Melissa Duarte², Niña Marada, Niña Charda, Niña Marada, Ni

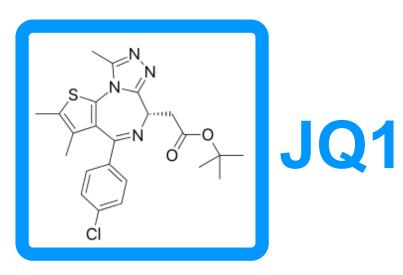
Colorectal Cancer

Loss of TRIM33 causes resistance to BET bromodomain inhibitors through MYC-and TGF-β-dependent mechanisms

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Leukemia

BET inhibitor resistance emerges from leukaemia stem cells



Ovarian Cancer

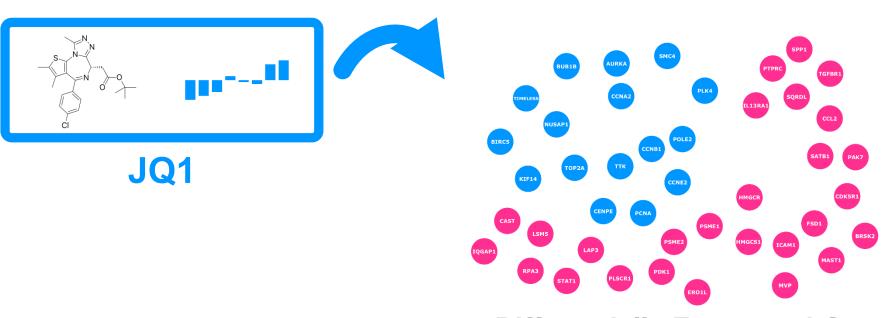
Resistance to BET Bromodomain Inhibitors Is Mediated by Kinome Reprogramming in Ovarian Cancer

Alison M. Kurimchak,¹ Claude Shelton,¹ Kelly E. Duncan,¹ Katherine J. Johnson,¹ Jennifer Brown,¹ Shane O'Brien,² Rahid Gabbasov,²-4 Lauren S. Fink,¹ Yusheng Li,¹ Nicole Lounsbury,³ Magid Abou-Gharbia,³ Wayne E. Childers,³ Denise C. Connolly,² Jonathan Chernoff,¹ Jeffrey R. Peterson,¹ and James S. Duncan¹.*

Identify a small molecule that would be effective in combination with JQ1

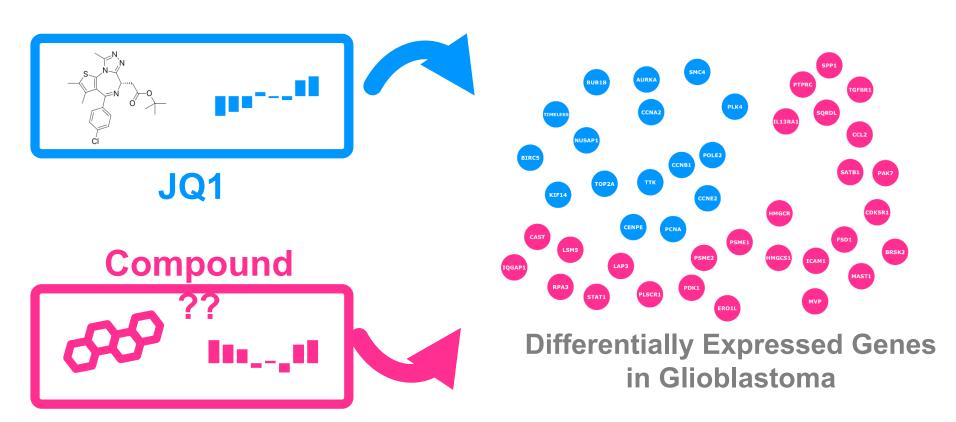


Identify compounds that induce orthogonal gene expression

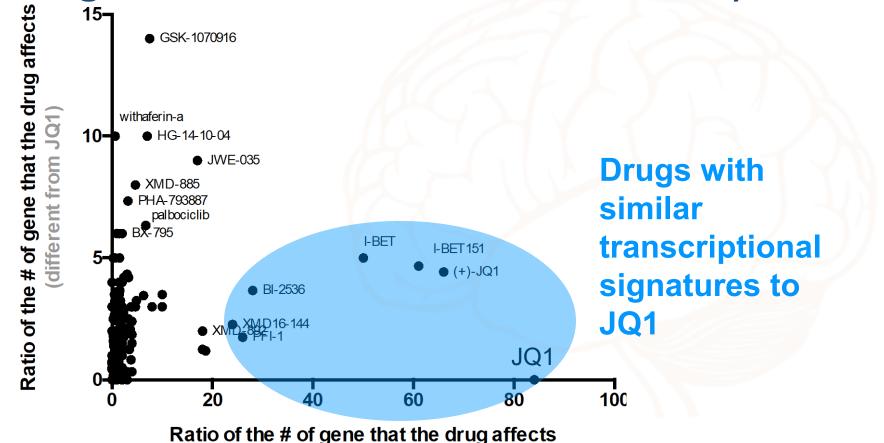


Differentially Expressed Genes in Glioblastoma

Identify compounds that induce orthogonal gene expression

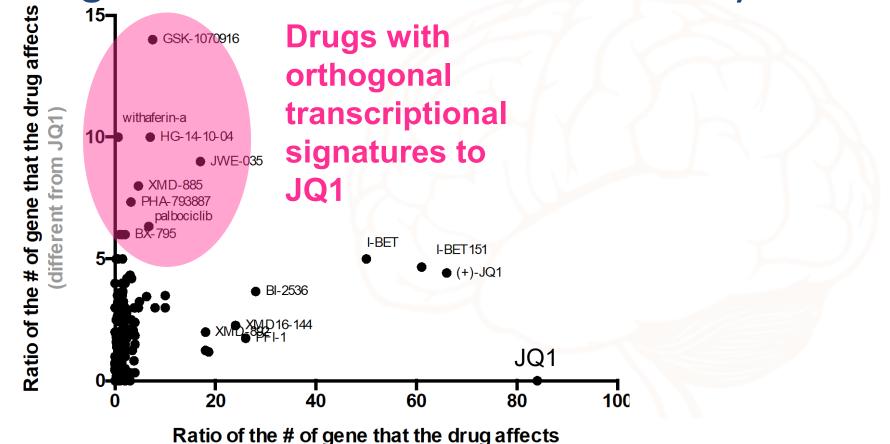


All Drugs were ranked based on similarity to JQ1



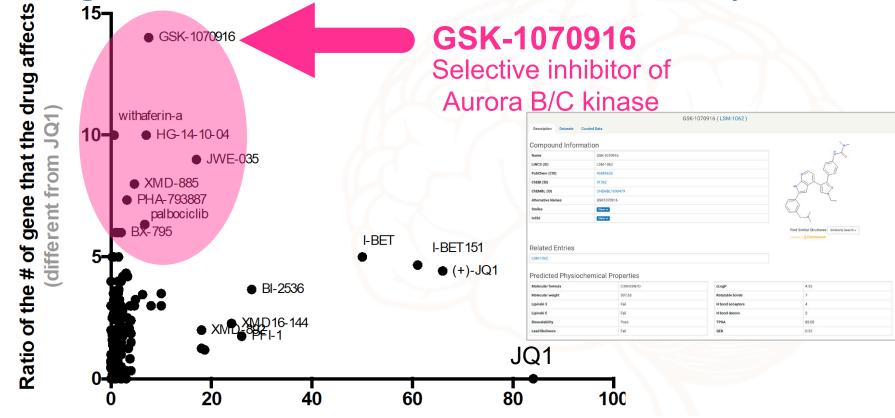
(similar to JQ1)

All Drugs were ranked based on similarity to JQ1



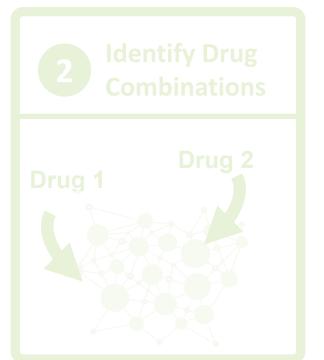
(similar to JQ1)

All Drugs were ranked based on similarity to JQ1



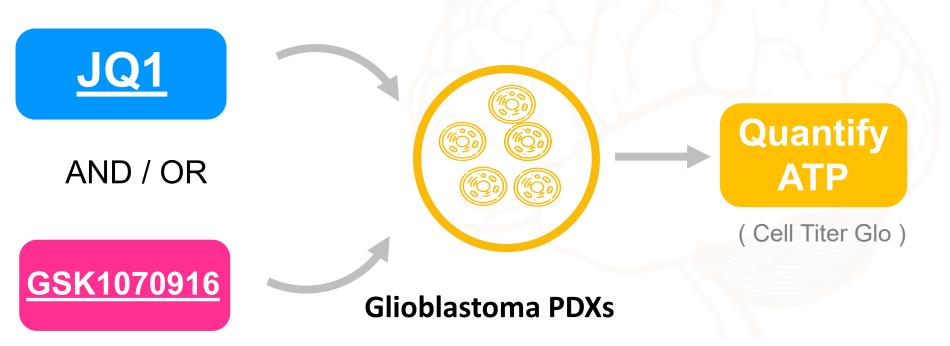
Ratio of the # of gene that the drug affects (similar to JQ1)



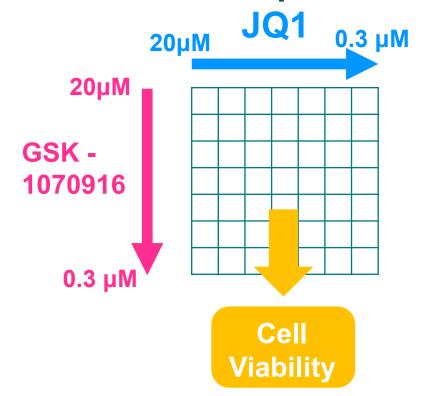


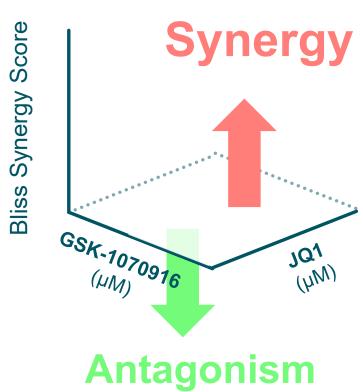


Treat Glioblastoma PDX cells with JQ1 and / or GSK-1070916



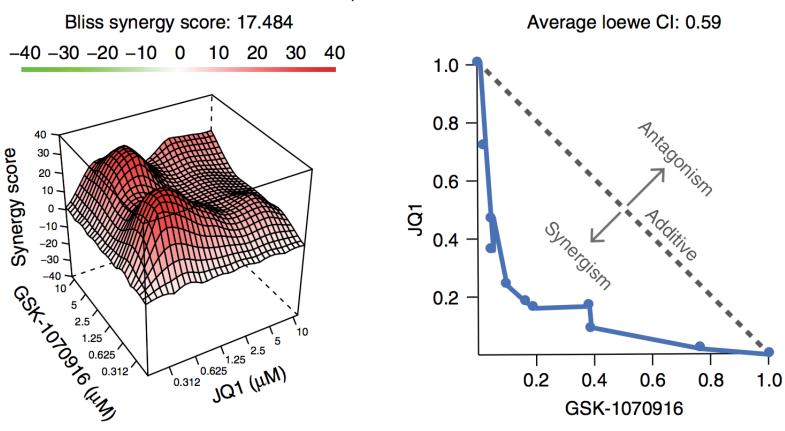
Bliss independence analysis of cell-viability combination response





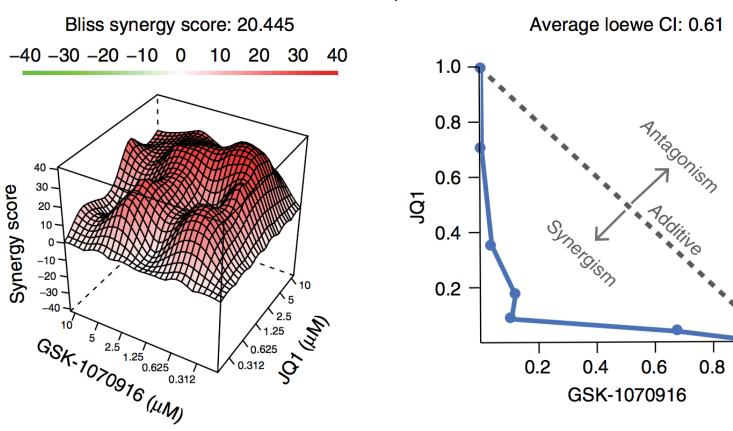
JQ1 + GSK-1070916 synergize in reducing cell viability

GBM39, CellTiter-Glo®



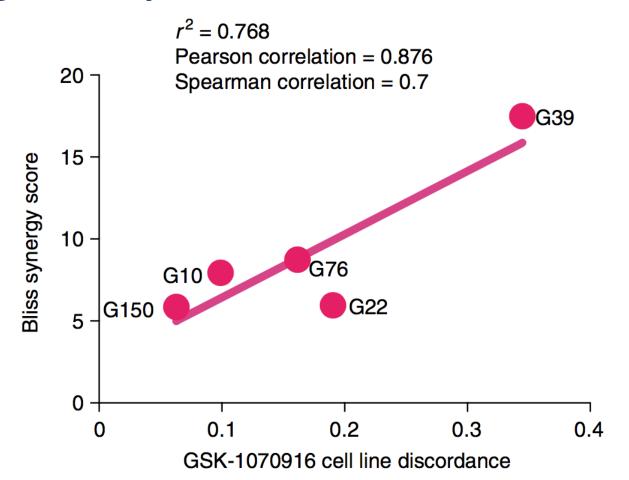
JQ1 + GSK-1070916 synergize in Glioblastoma cell death

GBM39, Caspase-Glo®



1.0

Synergistic response correlates with discordance





Introduction

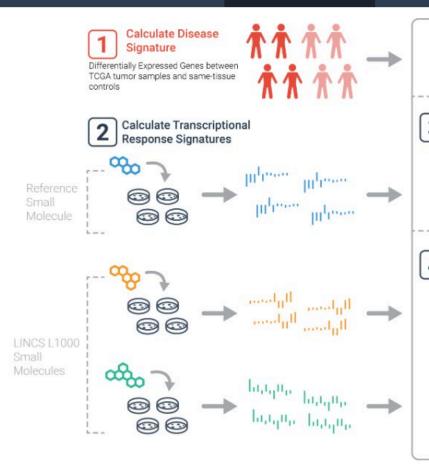
Tools

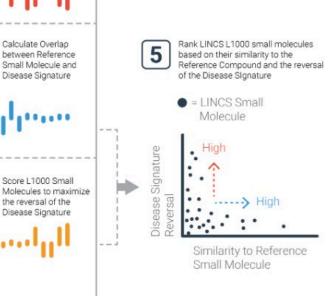
About

Calculate Overlap

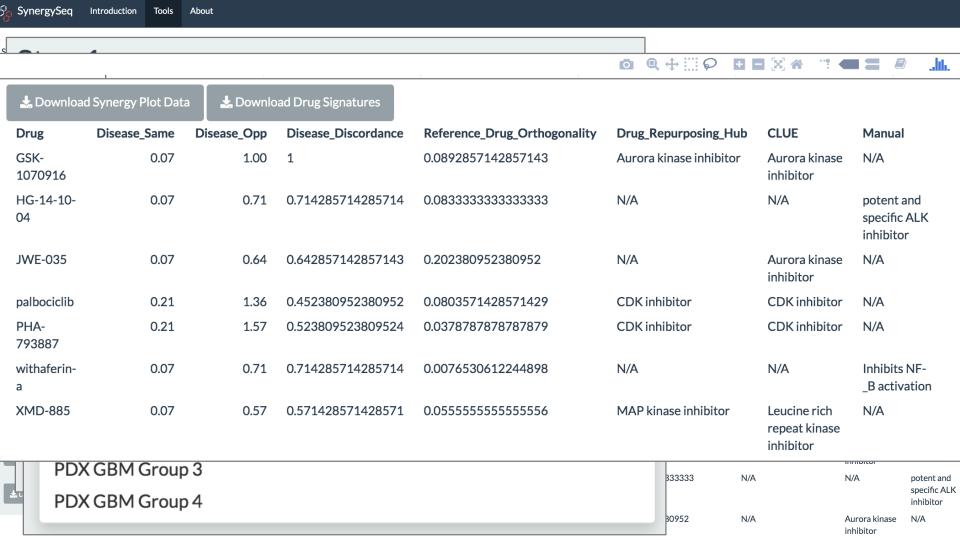
Disease Signature

the reversal of the Disease Signature





http://synergyseq.com/





Article | OPEN | Published: 14 December 2018

Drug and disease signature integration identifies synergistic combinations in glioblastoma

Vasileios Stathias, Anna M. Jermakowicz, Marie E. Maloof, Michele Forlin, Winston Walters, Robert K. Suter, Michael A. Durante, Sion L. Williams, J. William Harbour, Claude-Henry Volmar, Nicholas J. Lyons, Claes Wahlestedt, Regina M. Graham, Michael E. Ivan, Ricardo J. Komotar, Jann N. Sarkaria, Aravind Subramanian, Todd R. Golub, Stephan C. Schürer [™] & Nagi G. Ayad [™]

Abstract

Glioblastoma (GBM) is the most common primary adult brain tumor.

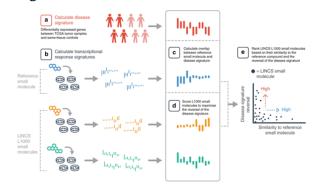








Fig. 1



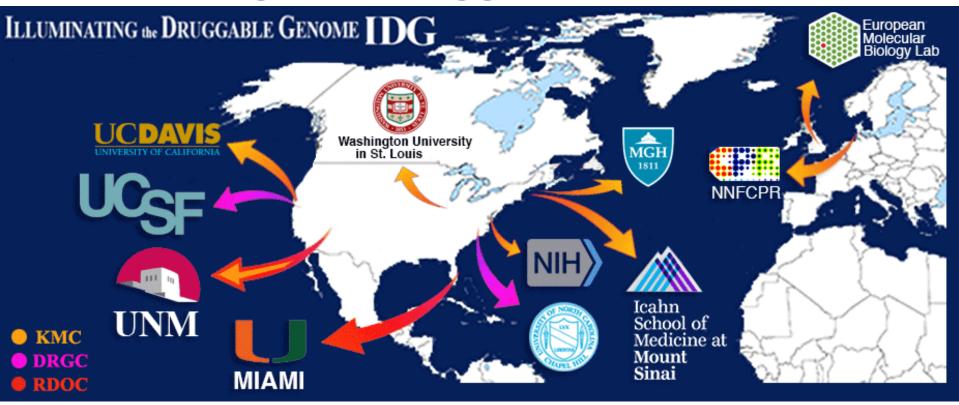
View in article

Full size image ≫

Take away (why did I show this)

- Use LINCS data (http://lincsportal.ccs.miami.edu)
- Representation of chemical structures by transcriptional consensus signatures
- Systems-level data driven approach vs targetdriven approach
- Current work: combining transcriptional signatures with with chemical structure-based signatures

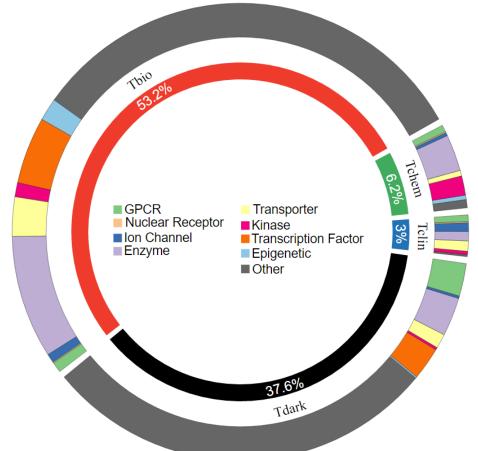
Illuminating the Druggable Genome (IDG)



http://druggablegenome.net

Target Development Level (TDL)





- Most protein classification schemes are based on structural and functional criteria.
- For therapeutic development, it is useful to understand how much and what types of data are available for a given protein, thereby highlighting well-studied and understudied targets.
- Proteins annotated as drug targets are **Tclin**
- Proteins for which potent small molecules are known are **Tchem**
- Proteins for which biology is better understood are **Tbio**
- Proteins that lack antibodies, publications or Gene RIFs are **Tdark**



Analysis | Published: 23 March 2018

Unexplored therapeutic opportunities in the human genome

Tudor I. Oprea, Cristian G. Bologa, Søren Brunak, Allen Campbell, Gregory N. Gan, Anna Gaulton, Shawn M. Gomez, Rajarshi Guha, Anne Hersey, Jayme Holmes, Ajit Jadhav, Lars Juhl Jensen, Gary L. Johnson, Anneli Karlson, Andrew R. Leach, Avi Ma'ayan, Anna Malovannaya, Subramani Mani, Stephen L. Mathias, Michael T. McManus, Terrence F. Meehan, Christian von Mering, Daniel Muthas, Dac-Trung Nguyen, John P. Overington, George Papadatos, Jun Qin, Christian Reich, Bryan L. Roth, Stephan C. Schürer, Anton Simeonov, Larry A. Sklar, Noel Southall, Susumu Tomita, Ilinca Tudose, Oleg Ursu, Dušica Vidović, Anna Waller, David Westergaard, Jeremy J. Yang & Gergely Zahoránszky-Köhalmi - Show fewer authors





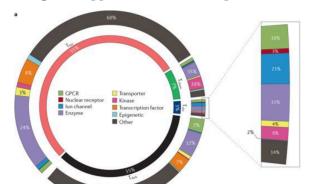
Figures

References

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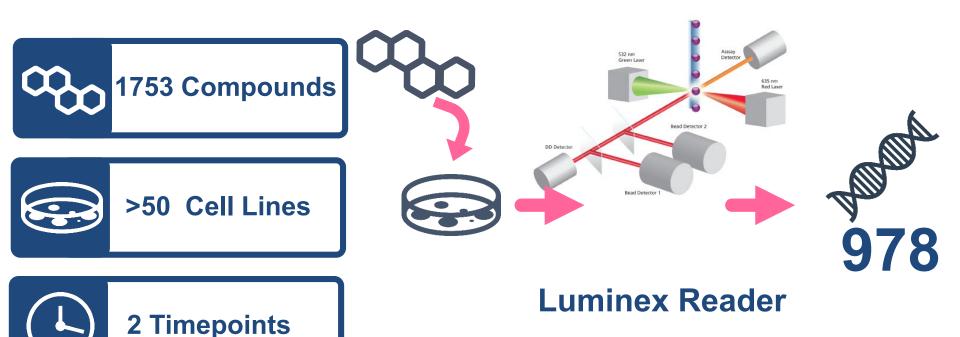
Sections

Figure 1: Target development level categories applied to the human proteome.



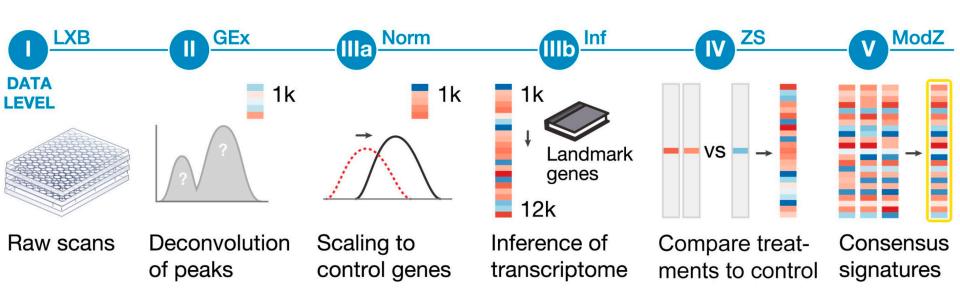
Additional slides

L1000 Gene Expression Platform



Aravind Subramanian et al, Cell, 2017

L1000 Computational Processing Pipeline



Aravind Subramanian et al, Cell, 2017

t-SNE Clustering of 4515 TCGA RNA-seq tumor samples

