Lecture 3

The MAX transcription factor (and why we want chemical probes)

February 22, 2024

Quarter-million genomes analyzed in NIH project could be 'hugely important' for identifying disease risks

All of Us finds many millions of new DNA variants as it moves toward 1 million participants from diverse backgrounds

19 FEB 2024 · 11:00 AM ET · BY JOCELYN KAISER

Science

That rich diversity in the project's first 245,388 whole genomes revealed more than 1 billion genetic variants, more than 275 million of which were novel. This expands by roughly 30% the catalog of known variants found by the UK Biobank and other whole genome databases, says Bick, corresponding author on the Nature paper.

Most of the 275 million new variants are rare, and many could prove harmless, which would allow clinicians to rule them out when searching for the gene behind a child's disease, for instance. Other variants may cause functional changes in a gene's protein that in turn trigger disease and could offer fresh targets for drug development.

From Lecture 1 (again): Therapeutically-driven probe discovery

assess tractability of emerging target candidates





chemical probe

cell lines and patient samples reveal list of disease genes test impact of disease genes in a physiologic settings

discover molecules that reverse impact of disease genes

The Cancer Dependency Map

interrogation of viability effects in cancer cell lines to map genetic dependencies



Pre-publication DATA RELEASES to enable the scientific community at depmap.org

As of 4/21/22: >2,000 cancer models 3913 genetic dependency screens 33 drug panels in sensitivity screens

Tsherniak et al., Cell, 170 (3): 564-576 (2017) McDonald et al., Cell, 170 (3): 577-592 (2017) Vazquez & Boehm, Mol Sys Bio, 16 (7): e9757 (2020) Dharia et al., Nat Genet, 53 (4): 529-538 (2021)

example query: multiple myeloma

Dependencies enriched in Multiple Myeloma

				🛃 😧 Show/H	de Columns or disappear
Туре 💭	Gene/Com	pound	Dataset	T-Statistic	P-Value
gene	IRF4	*	CRISPR (DepMap 21Q4 Public+Score, Chronos)	-24	1.17E-101
gene	PRDM1	*	CRISPR (DepMap 21Q4 Public+Score, Chronos)	-23.2	3.48E-96
gene	PRDM1	*	CRISPR (DepMap 21Q4 Public, Chronos)	-23.5	6.94E-96
gene	IRF4	*	CRISPR (DepMap 21Q4 Public, Chronos)	-22.4	2.5E-89
gene	IRF4	*	CRISPR (Project Score, Chronos)	-24.4	1.14E-74
gene	IRF4	*	RNAi (Achilles+DRIVE+Marcotte, DEMETER2)	-20.4	7.31E-73
gene	PIM2		CRISPR (DepMap 21Q4 Public+Score, Chronos)	-19.2	9.45E-71
gene	POU2AF1	*	CRISPR (DepMap 21Q4 Public+Score, Chronos)	-18.6	8.99E-67
gene	PIM2		CRISPR (DepMap 21Q4 Public, Chronos)	-18.8	1.16E-66
gene	IRF4	*	CRISPR (Project Score, CERES)	-22.2	1.43E-66
gene	NFKB1	*	CRISPR (DepMap 21Q4 Public, Chronos)	-18.6	6.92E-66
gene	POU2AF1	*	CRISPR (DepMap 21Q4 Public, Chronos)	-18.6	1.56E-65
gene	MEF2C	*	CRISPR (DepMap 21Q4 Public+Score, Chronos)	-18.1	1.11E-63
gene	NFKB1	*	CRISPR (DepMap 21Q4 Public+Score, Chronos)	-17.8	2.44E-62
gene	HERPUD1		CRISPR (DepMap 21Q4 Public+Score, Chronos)	-17.8	5.97E-62
gene	IRF4	*	RNAi (DRIVE, DEMETER2)	-19.3	2.52E-58
gene	HERPUD1		CRISPR (DepMap 21Q4 Public, Chronos)	-16.8	3.31E-55
gene	SMAD7		CRISPR (DepMap 21Q4 Public+Score, Chronos)	-16.6	5.65E-55
gene	TCF3	*	CRISPR (DepMap 21Q4 Public+Score, Chronos)	-15.9	4.48E-51
gene	TCF3	*	CRISPR (DepMap 21Q4 Public, Chronos)	-16	4.96E-51

IRF4 interferon regulatory factor 4

Overview Dependency

Characterization Description

All

2

3

4

5 6

7

All

3

example query: IRF4



Dependency Score: Outcome from DEMETER2 or CERES. A lower score means that a gene is more likely to be dependent in a given cell line. A score of 0 is equivalent to a gene that is not essential whereas a score of -1 corresponds to the median of all common essential genes.

Strongly Selective: A gene whose dependency is at least 100 times more likely to have been sampled from a skewed distribution than a normal distribution (i.e. skewed-LRT value > 100).

View more

The protein encoded by this gene belongs to the IRF (interferon regulatory factor) family of transcription factors, characterized by an unique tryptophan pentad repeat DNA-binding domain. The IRFs are View more

Search external sites for IRF4

- PubMed (996 entries)
- GeneCards
- GTEx
- NCBI

Enriched Lineages CRISPR 1 . -2 0 and the second second -2 -2.0-1.5-1.0-0.5 0.0 0.5 Dependency Score (CERES) 1. Multiple Myeloma (4.4e-79) 2. Haematopoietic And Lymphoid (3.4e-39) 3. Solid (4.5e-38) 4. Lymphoma (4.0e-15) 5. T-cell lymphoma Other (9.5e-12) 6. ALCL (9.9e-07) 7. Melanoma (1.3e-04) **RNAi** 1 . 2 ' -1.5-0.5-1.00.0 0.5 Dependency Score (DEMETER2)

1. Multiple Myeloma (1.3e-72) 2. Haematopoietic And Lymphoid (3.3e-28) 3. Solid (6.5e-21)





Number of Selective Gene Dependencies Identified Cancer Dependency Screening by Molecular Function Data from Broad Institute Cancer DepMap Project 800 600 Total 400 200 Membrane lattic vere extracellular matrix brodein isonnerase kinase Cell adhesion on the cule defense innunity of ofen Ruy binding protein transfercanier protein * DN4 binding Drodein signaling molecule ^{transcription factor} ""um-binding profein Cell Junction Drotein phosphatese Iransferase L transporter hydroldse Drofease

36 out of the top 80 dependencies are TFs





https://depmap.org/portal/depmap/



Can we build general and systematic platforms for developing chemical probes for transcriptional regulators?

Transcription Profile

expression gene **Metastatic Tumors Pre-Cancerous** Tumor Cell Stage

Can we tune or reprogram dysregulated gene expression programs and impact cell state?

MYC family of transcription factors

'master regulators' of broad cellular processes



Secondary RNA amplification

MYC

accumulates in promoter regions and amplifies transcription when overexpressed in cancer



Lin et al., Cell, 151, 56-67 (2012); Nie et al, Cell 151, 68-79 (2012)

MYC

accumulates in promoter regions and amplifies transcription when overexpressed in cancer



'silver bullet' drug

Lin et al., Cell, 151, 56-67 (2012); Nie et al, Cell 151, 68-79 (2012)

Cancers dysregulate MYC by increasing its expression



in typical cells, steady state MYC levels regulate general housekeeping functions

MYC can be transiently upregulated in typical cells (e.g. during wound healing)

tumor cells need persistently upregulated MYC at super physiologic levels to drive tumor-specific oncogenes

Adapted from Wolf et al., Trends Cell Biol, 25, 241-248 (2015)

Oncogenic levels of MYC regulate all hallmarks of cancer



MYC expression in haploinsufficient mice

amelioration of age-associated phenotypes





MYC is an obstinate therapeutic target



many protein-protein interactions

unstructured domains no traditional binding pockets large buried interface

Myc/Max/Mxd Network

alternative paths to modulating amplified Myc-driven transcription in cancer



Adapted from Diolaiti, McFerrin, Carroll, Eisenmann, Biochimica et Biophysica Acta 1849, 484-500 (2015)

MAX as a target: alter heterodimer/homodimer dynamics



MAX: Myc-Associated factor X



bloactive Compounds	Tes
Druggable Structure	No
Druggable by Ligand Based	No
Assessment	
Enzyme	No







splice variants

AlphaFold (predicted)

crystal structure with Myc basic helix loop helix leucine zipper bHLH-LZ

MAX: Myc-Associated factor X

Cellular localization: primarily nuclear high levels in brain, heart, lung Tissue specificity: low levels in liver, kidney, skeletal muscle acetylation (localization) Post-translational mod: phosphorylation (stability) mutated in pheochromocytoma Diseases: mutated in small cell lung cancers potential tumor suppressor role in 'neuroendocrine' tumors, which are tumors that form from cells that

release hormones into the blood in response to signals from the nervous system

SMM screens: purified Max transcription factor



117 assay positives

Reporter gene assays: putative Max binders modulate Myc-driven transcription



Cell viability assays: Are Myc or Max required?



P493-6 Myc off

>50 µM

>50 µM

	P493-6	PC12
Max	3	1

Cell viability assays: Are Myc or Max required?



Conditional cellular models of MYC expression



Concentration (µM)

Imaging of biomarkers: conditional vs. chemical modulation

modulating Myc in an engineered osteosarcoma model



Anja Deutzmann, Felsher Lab Stanford

scale bar = 40 μm

Does the probe antagonize the Myc/Max heterodimer?



Electrophoretic Mobility Shift Assay (EMSA) aka Gel Shift Assay



Does the probe antagonize the Myc/Max heterodimer?



EMSA

Does the probe stabilize the Max/Max homodimer?



0

Western blots: KI-MS2-008 alters Myc protein levels







rescue experiment with 10 µM proteasome inhibitor MG132

Myc protein stability is regulated by the ubiquitin-proteasome system



Proteomics by mass spec: KI-MS2-008 decreases MYC

proteome-wide measurements (10 µM KI-MS2-008)



KI-MS2-008 – a mixed mechanism probe?

competition for DNA binding + destabilization of MYC



CHIP-Seq or CHromatin ImmunoPrecipitation coupled to Sequencing is a protein bound to a piece of DNA or not?



CHIP-Seq

KI-MS2-008 perturbs binding of Myc and Max at promoters of *MYC*-occupied genes



Struntz et al., Cell Chem Biol, 26, 711-723 (2019)

In vivo studies: KI-MS2-008 modulates tumor volume in Myc-dependent mouse models of cancer

T-cell acute lymphoblastic leukemia blood cancer







vehicle

MS2-008



0.24 mg/kg subcutaneous administration 5d on/2d off cycles

radiance (p/s/cm²/sr)

Cell Chemical Biology

Stabilization of the Max Homodimer with a Small Molecule Attenuates Myc-Driven Transcription

Graphical Abstract



Highlights

- KI-MS2-008 is a Max-binding small molecule that attenuates Myc-driven transcription
- The compound stabilizes the Max homodimer
- Effects on DNA occupancy and the transcriptome resemble loss of Myc
- Treatment with KI-MS2-008 exhibits efficacy in cellular and murine cancer models

Authors

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In Brief

Myc/Max-mediated transcription is deregulated in most of human cancers. Struntz et al. discovered a small molecule that stabilizes the Max homodimer and attenuates Myc-driven transcription with efficacy in cellular and murine cancer models. This discovery reinforces an alternative Myc-targeting strategy and could inform development of compounds to treat Myc-dependent cancers.





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For future drug hunters – Lipinski's Rule of 5

1997 - is a rule of thumb to evaluate 'drug-likeness' or determine if a chemical compound with a certain pharmacological or biological activity has chemical properties and physical properties that would make it a likely orally active drug in humans (Chris Lipinski, Pfizer)



MW < 500 Da CLogP < 5 H-bond donor < 5 H-bond acceptor < 10



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MW < 500 Da CLogP < 5 H-bond donor < 5 H-bond acceptor < 10





MW = 536.65 ClogP = 6.38 H-bond donor = 0 H-bond acceptor = 4





Can we screen MAX against a new chemical library and find compounds with better physicochemical properties?

Can we find MAX binders with different modes of action?

Upcoming Lectures

- 2/8/24 Lecture 1 Intro to chemical biology: small molecules, probes, and screens
- 2/13/24 No Lecture Snow Day
- 2/15/24 Lecture 2 Small Molecule Microarrays
- 2/20/24 No Lecture
- 2/22/24 Lecture 3 Our protein target MAX
- 2/27/24 Lecture 4 Quantitative evaluation of protein-ligand interactions
- 2/29/24 Lecture 5 KB-0742: A Phase 2 clinical candidate discovered by SMMs
- 3/5/23 Lecture 6 Wrap up discussion for Mod 1 experiments and report