



Design of individual gRNAs and Genome-wide Libraries of gRNAs

Workshop 4

Keystone Symposium, Precision Genome Engineering
Breckenridge, CO, January 8 – 12, 2017

John Doench
Broad Institute of MIT and Harvard



I want the best tool for the job!

...but there are a lot of jobs out there!



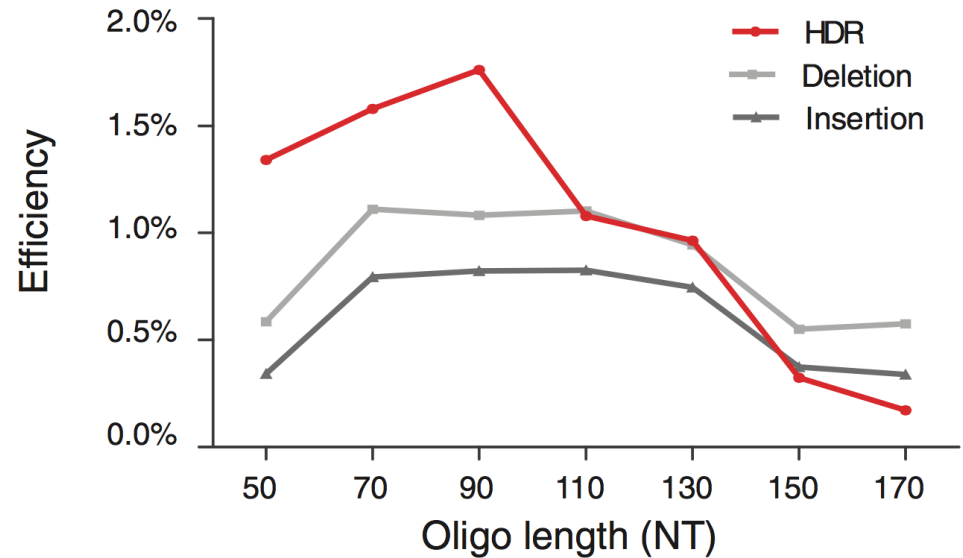
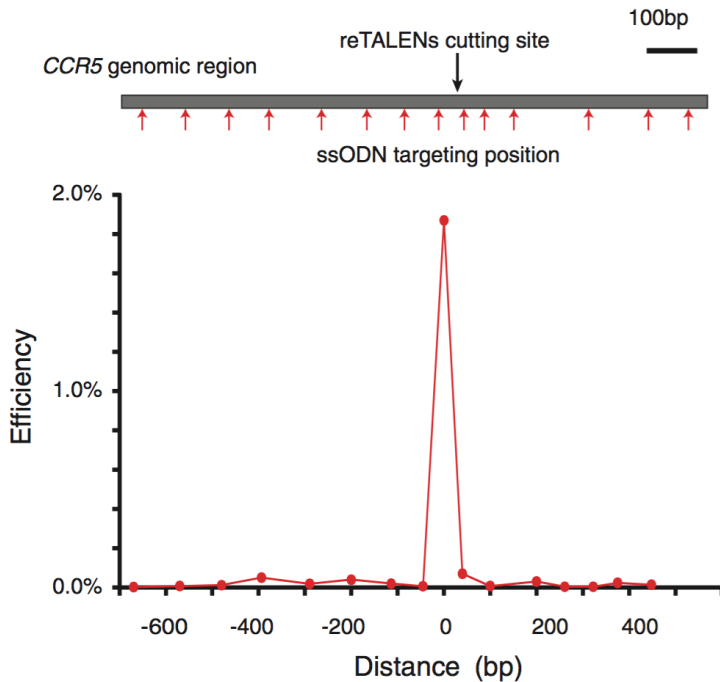
- Purpose
 - Gene editing or gene disruption
- Time
 - Long-term model or short-term investigation
- Scale
 - Study your favorite gene or genome-wide screen
- Delivery of Cas9
 - Lentivirus, AAV, plasmid, mRNA, protein
- Delivery of guide RNA
 - Lentivirus, AAV, plasmid, cr + trRNA, sgRNA

“The best” will vary based on what you’re doing

Gene editing



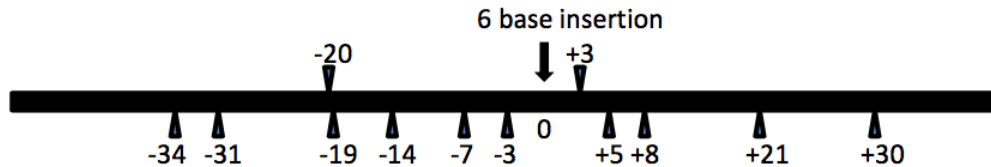
- Guide design is largely irrelevant, because you have few choices
 - Want the cut site to be as close as possible to the intended mutation



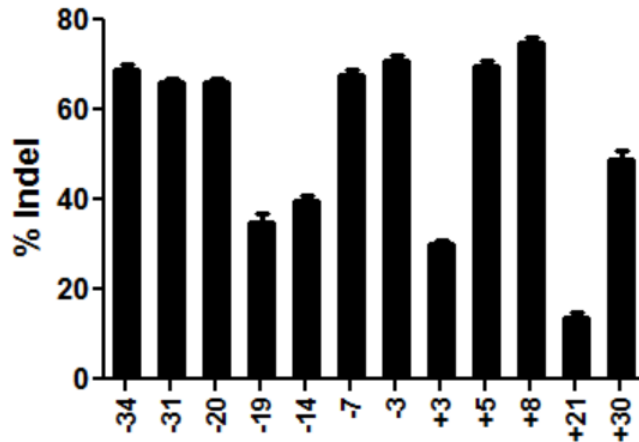
Gene editing



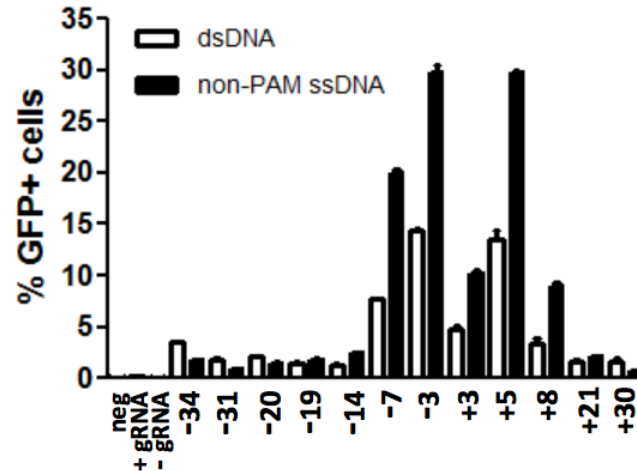
(A) Available gRNAs flanking the insertion site



(B) gRNA cleavage efficiency



(C) dsDNA or ssDNA donors

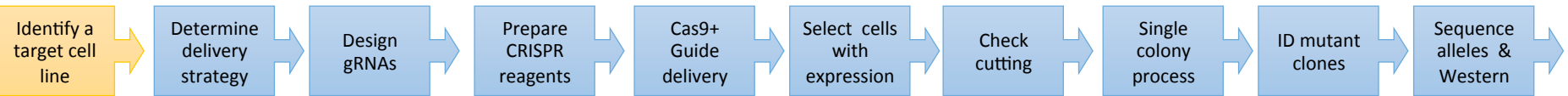


What are the choices for gene editing?



- Large edit that can carry a selectable marker (i.e. knock-in) or single-base change?
- How well do your cells single cell clone?
 - If poorly, can you get away without single cell cloning with a smart assay?
- Modified RNA components?
- Modified Cas9 enzymes?
- NHEJ-inhibitors?
- Delivery delivery delivery

CRISPR workflow for cell line engineering: Selecting a cell line



- HEK293, HCT116, and K562 cell lines are the most commonly used and represent the simpler cell lines to edit
- Choose a cell line that **best captures your biology** or question of interest and then build a CRISPR delivery solution around it
- **Characterize your cell lines!**
 - Copy number, best method of transfection, capable of single cell expansion, SNPs in gene of interest
- Unless a primary cells such as fibroblast can undergo 20-30 doublings without senescing, they are usually *not* amenable
 - Consider iPSc or immortalization
 - ES/iPSc editing requires special workflows to maintain pluripotency throughout the editing process

Make sure you fingerprint/ID and mycoplasma test your cell lines before starting a genome editing workflow –
garbage in, garbage out

For gene editing

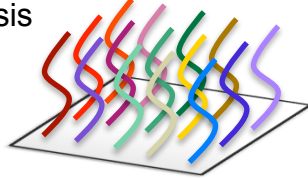


- Design is more about *experimental* design rather than choosing the right guide RNA, because choices are so limiting. So, workflow:
 1. Optimize delivery to your system using pre-validated reagents (e.g. Cas9 that you know works, guide that you know works, etc.)
 - Also use these experiments to monitor cell viability
 2. Develop an assay to assess successful gene editing on the pool of cells (e.g. PCR & MiSeq)
 3. Test several combinations of guides and repair templates and used pooled assay from step 2
 4. Move forward with combos that work best, preferably with two different guides identified in step 3

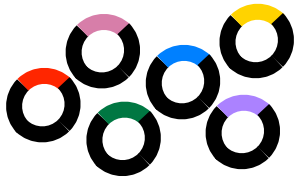
Genome-wide Pooled Screening









1. Oligonucleotide Synthesis

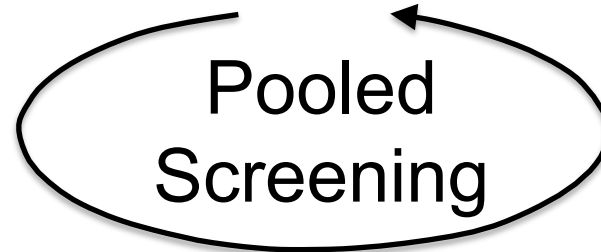


2. Plasmid Pool

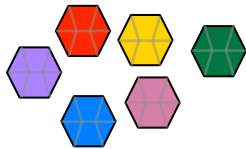


sgRNA	Init.	Pos.	Neg.
	16	0	21
	19	100	19
	17	0	0
	18	0	26
	14	0	18
	16	0	16

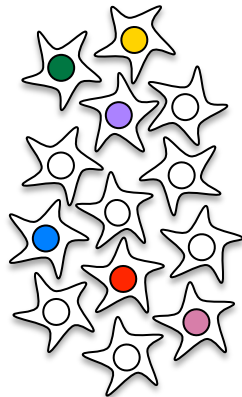
8. Matrix of sgRNA Abundance



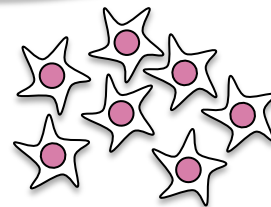
3. Virus Pool



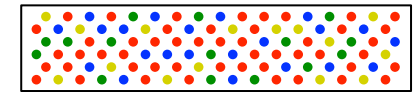
4. Pooled Infection



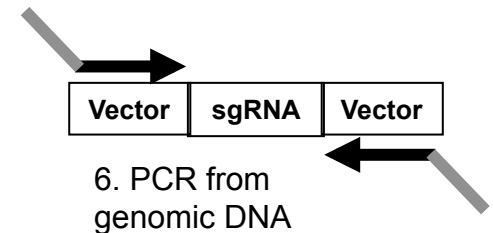
5. Positive Selection



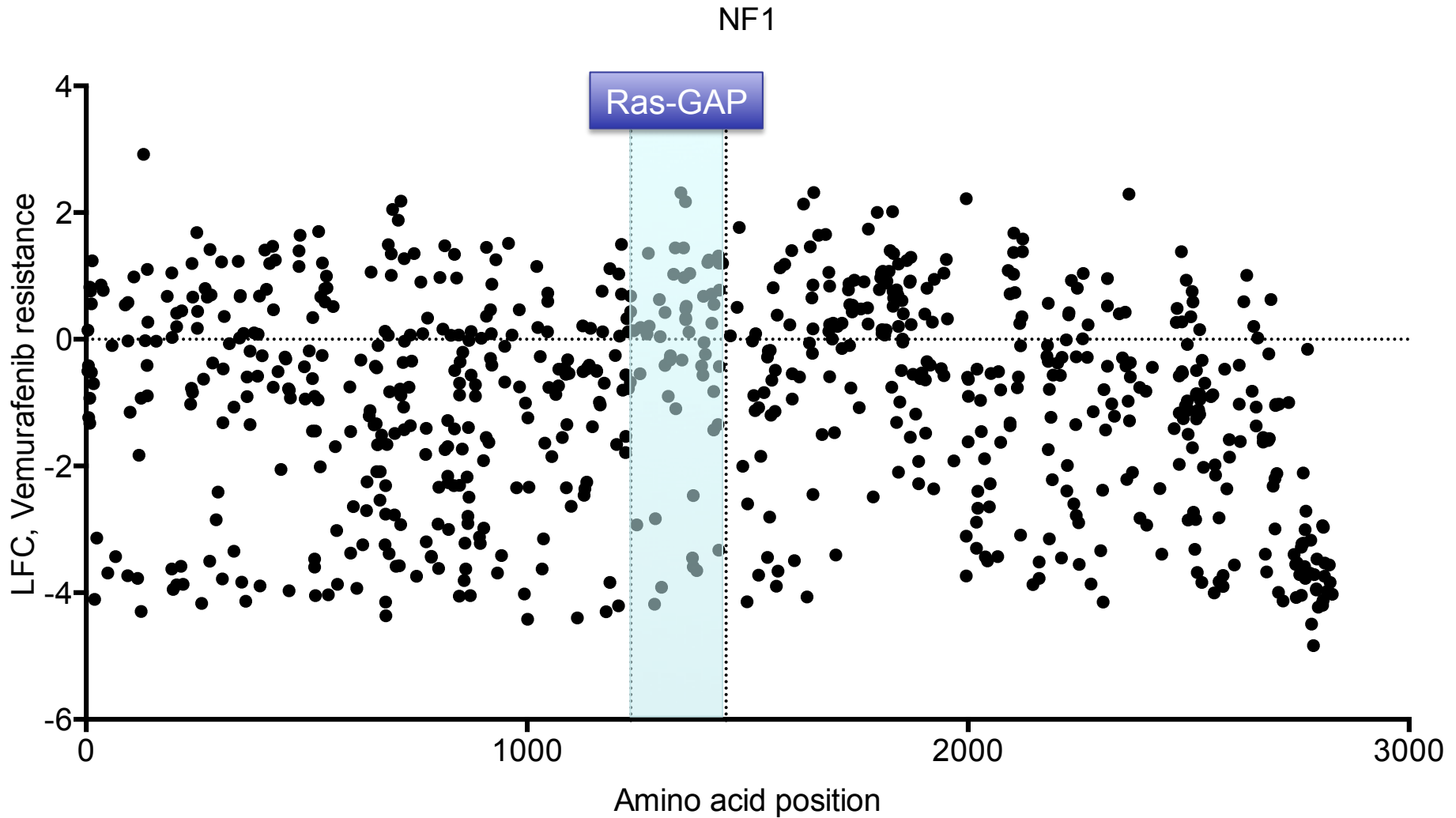
5. Negative Selection



7. NextGen Sequencing



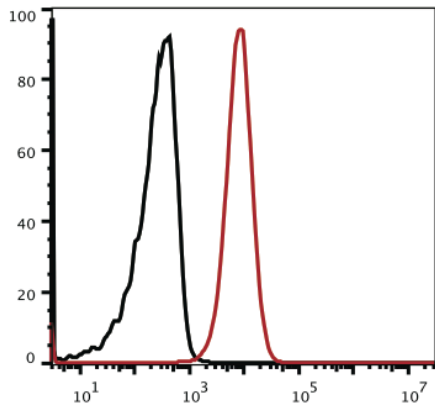
Variability in sgRNA efficacy



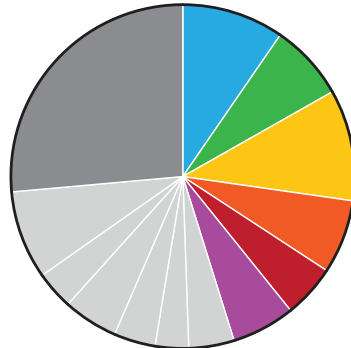
FACS based-readout: clear isolation of negative populations



Cd45

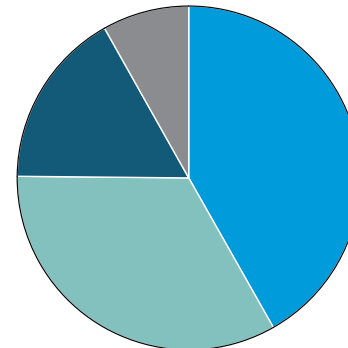


Mouse



Total=6085

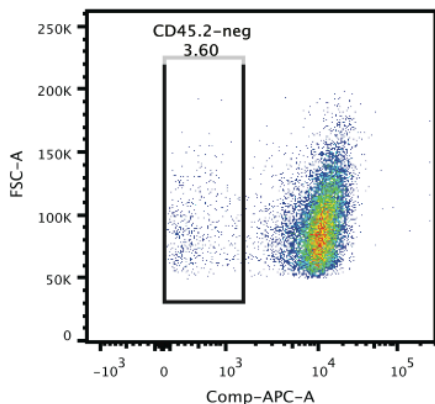
Human



Total=1151

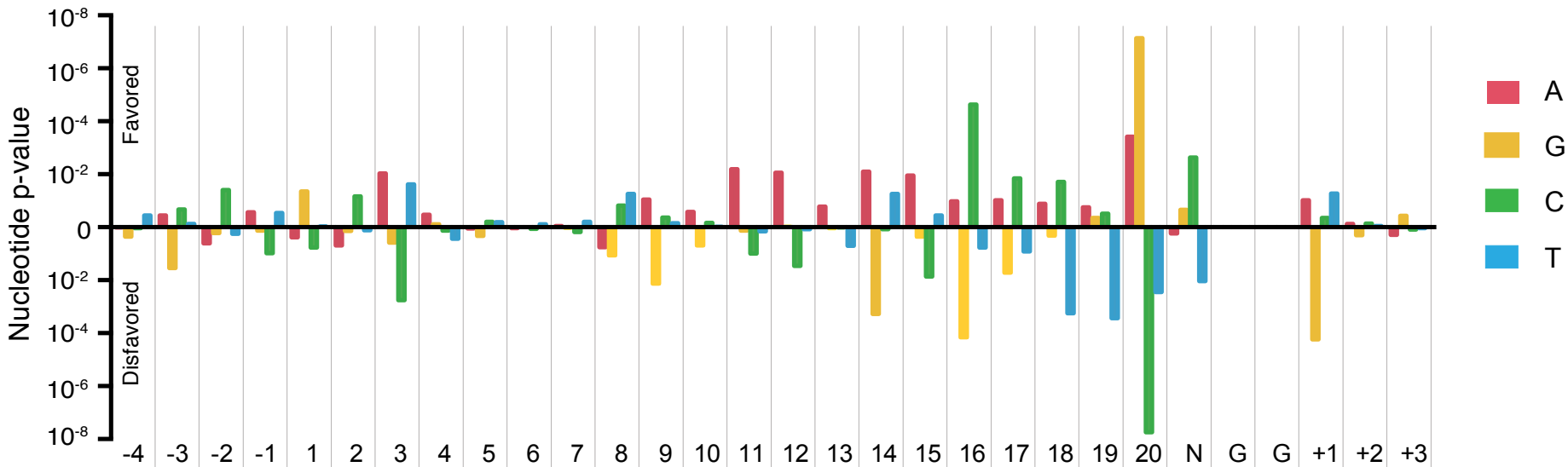
Cd45
Cd28
Cd43
Cd5
H2-K
Thy1
Cd3e
Cd2
Cd53
H2-D
CtlA-4
Cd4
Negative Control

CD13
CD15
CD33
Negative Control



- Create all possible sgRNAs against panel of cell surface markers
- Isolate cells that are negative for these markers
- Sequence to determine the frequency of each sgRNA
- **Enriched** in the marker-negative population are **active** sgRNAs
- Use of positive selection assays to determine rules means we are *not* selecting for promiscuity, in contrast to rules based on cell death

Nucleotide frequencies indicate preferences for highly-active sgRNAs

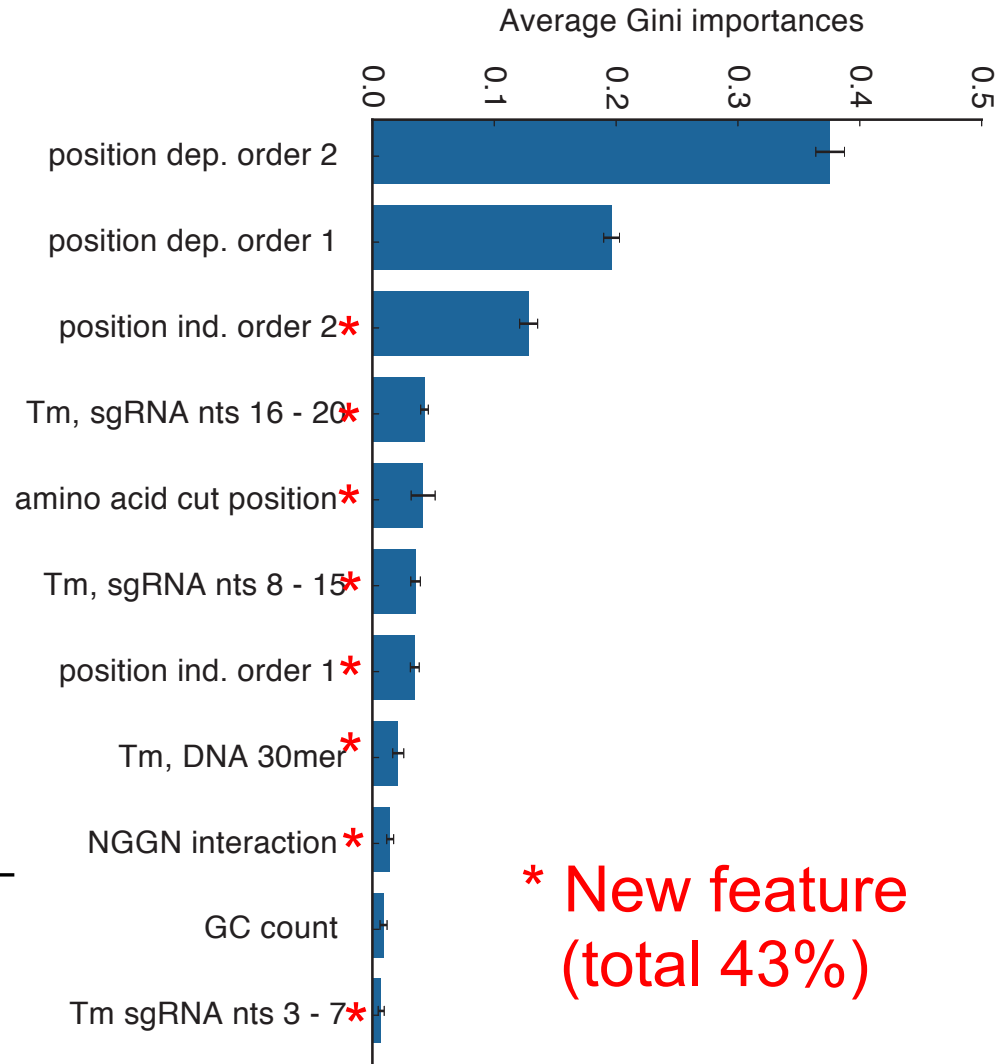
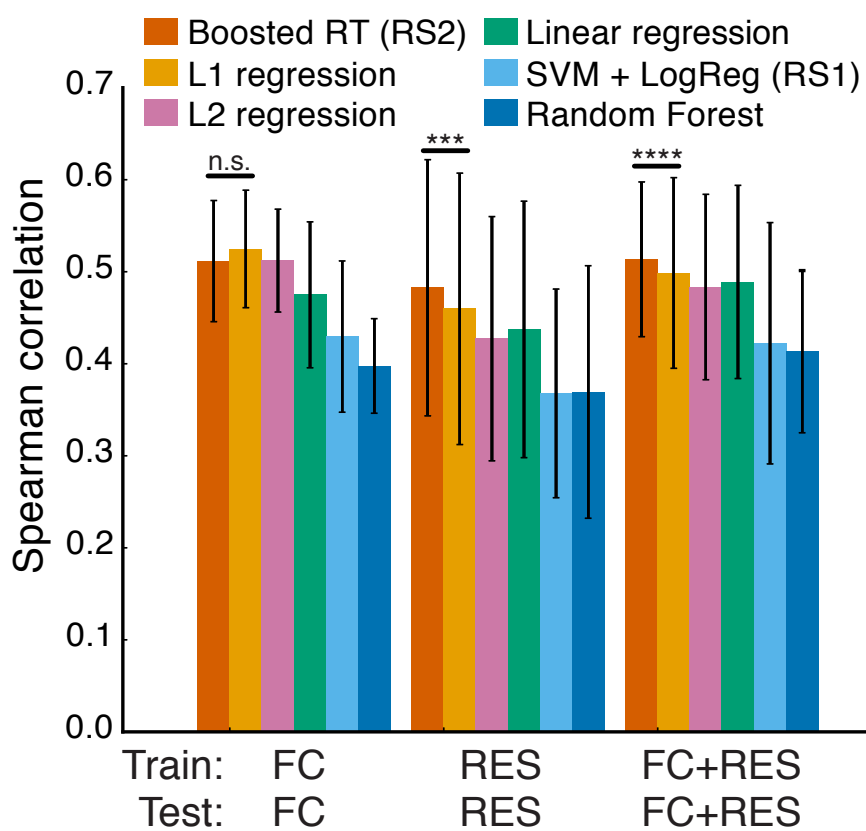


- For each gene, compare sequence content for 20% most-active sgRNAs to the remainder
- Linear weights of nucleotide features to build a classifier to predict sgRNA activity
- <http://portals.broadinstitute.org/gpp/public/analysis-tools/sgrna-design>

From Rule Set 1 to Rule Set 2: More data, better computation



with Jennifer Listgarten, Nicolo Fusì,
Microsoft Research












* New feature
(total 43%)

Unbiased comparison of prediction algorithms

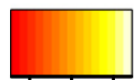


Guides transcribed in cells from a U6 promoter

Datasets

Wang/Xu HL60 (2076)		0.616	0.343	0.486	0.321	0.246	0.201	0.485
Doench 2014 Mouse-EL4 (951)		0.427	0.577	0.400	0.403	0.369	0.156	0.700
Koike-Yusa/Xu 1 M-ESC (907)		0.281	0.221	0.306	0.12	0.119	0.094	0.367
Chari 293T (1234)		0.310	0.246	0.286	0.457	0.308	0.123	0.381
Doench 2016 A375 (2333)		0.265	0.266	0.287	0.245	0.164	0.144	0.540
Hart Repl2Lib1 Hct116 (4239)		0.307	0.288	0.292	0.208	0.232	0.159	0.384
Gandhi Electrop. <i>Ciona</i> (72)		0.298	0.245	0.150	0.248	0.112	0.354	0.419
Farboud <i>C. elegans</i> (50)		0.476	0.301	0.545	0.602	0.400	0.177	0.541
Ren <i>Drosophila</i> (39)		0.313	0.178	0.225	0.152	-0.158	-0.347	0.131

Color Key



-0.5 0.5
Value

Wang Score

Doench Score

Xu Score

Chari Score

Wong Score

Moreno-Matos Score

Fusi/Doench Score

Predictions

Haeussler et al.,
Genome Biology, 2016



How the sgRNA is made / delivered may affect prediction algorithm accuracy



Guides transcribed in cells from a U6 promoter

K562 LacZ Rank (24)		-0.037	0.320
U2OS/MEF/C6 T7Endo (52)		-0.178	0.415

Guides transcribed *in vitro* from a T7 promoter

Mouse in vivo (30)		0.426	-0.304
Zebrafish Seq. (103)		0.240	0.103

Moreno-Mateos Score
Fusi/Doench Score

A selection of design sites



- Broad-GPP (for designing sgRNAs)
 - <http://portals.broadinstitute.org/gpp/public/analysis-tools/sgrna-design>
- CRISPOR (implements multiple scoring schemes)
 - <http://crispor.tefor.net>
- GenomeCRISPR (to search existing data)
 - <http://genomecrispr.dkfz.de/#!/>
- Companies
 - Desktop Genetics, Benchling, GE (Dharmacon), many others

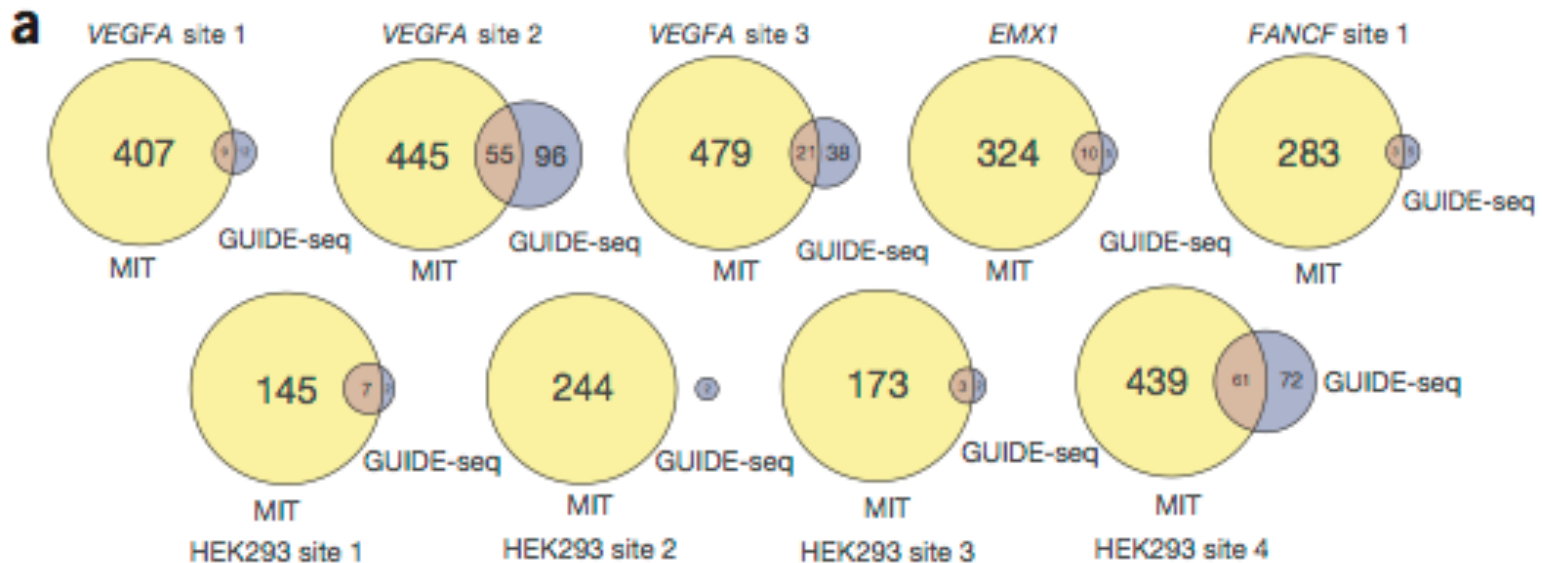


Don't use ANY tool unless you know what its limitations are and what purpose it was built for

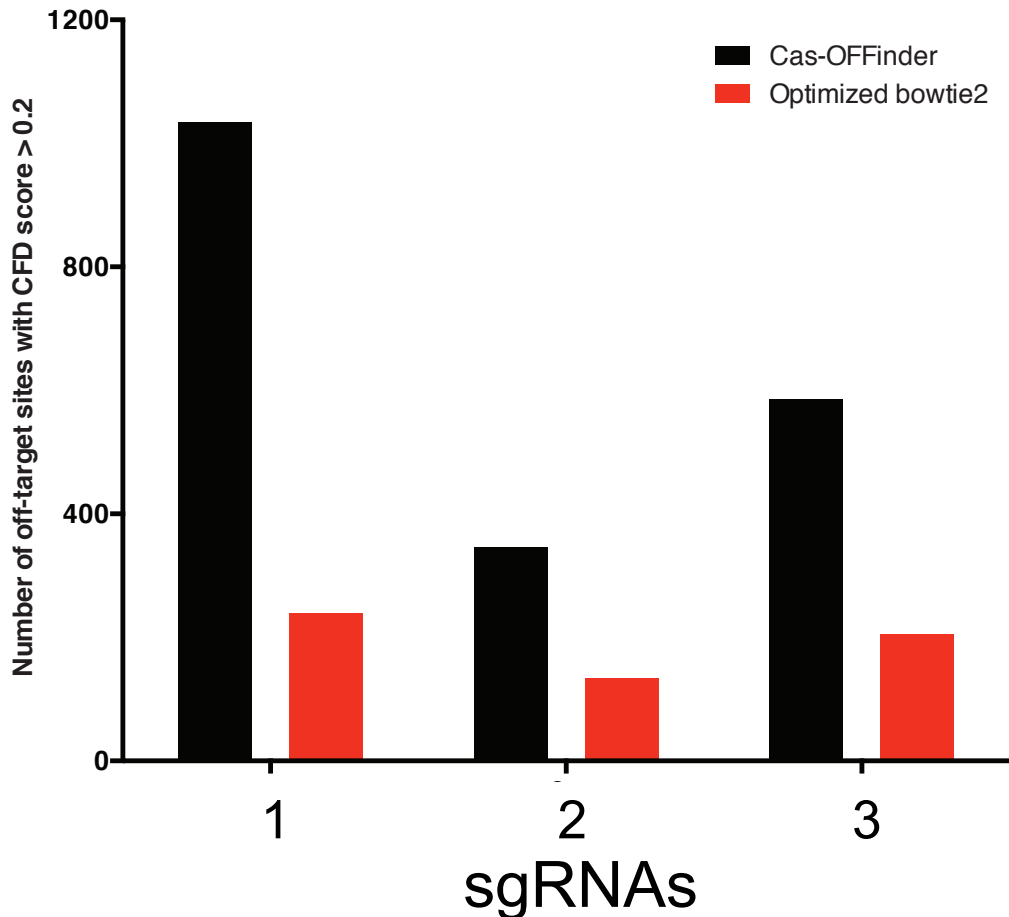
GUIDE-seq: unbiased detection of off-target sites



“Having established the efficacy of GUIDE-seq, we next performed direct comparisons of our method with two popular computational programs for predicting off-target mutation sites: the MIT CRISPR Design Tool and the E-CRISP software. Both of these programs identify potential off-target sites based on ‘rules’ about mismatch number and position. In direct comparisons, we discovered that **neither program identified the vast majority of off-target sites** found by GUIDE-seq for the nine RGNs.”



Often-used search algorithms miss potential off-target sites



Don't use design tools & websites that haven't been updated in several years, this field moves fast!

Genome-wide libraries

Available via Addgene as plasmid AND lentivirus



Human library
Avana



Mouse library
Asiago

From Brunello to Zinfandel – strategy of pooled oligo synthesis, pooled clone production, and pooled screening allows for *rapid, inexpensive* cycles of innovation



Human library
Brunello



Mouse library
Brie

Comprehensive resource for pooled CRISPR libraries: Addgene



Find Pooled Libraries of Interest

Click on different properties to create a custom filtered list of the CRISPR pooled libraries in Addgene's collection.

Library Type

- Knockout
- Activation
- Inhibition

Species

- Human
- Mouse
- Fly
- T. gondii*



Library Size

- Genome-wide
- Subpool

Viral Prep Available

- Yes
- No

Clear Filters

Name	ID	Library Type	Species	PI	Lentiviral Generation	gRNAs per gene	Total gRNAs
Activity-optimized genome-wide library	1000000067	Knockout	Human	Sabatini and Lander	3rd	10	178,896
Broad GPP genome-wide Brunello	73179 (1 plasmid) 73178 (2 plasmid) 	Knockout	Human	Doench and Root	3rd	4	76,441
Broad GPP genome-wide Brie	73632 (1 plasmid) 73633 (2 plasmid) 	Knockout	Mouse	Doench and Root	3rd	4	78,637
Broad GPP kinome Brunello	75314, 75315 (1 plasmid) 75312, 75313 (2 plasmid)	Knockout	Human	Doench and Root	3rd	4	3,052
Broad GPP kinome Brie	75317 (1 plasmid) 75316 (2 plasmid)	Knockout	Mouse	Doench and Root	3rd	4	2,852
CRISPRa	Discontinued	Activation	Human	Weissman	3rd	10	198,810



Are we doing any better? sgRNA sets to assess library performance



Article



molecular
systems
biology

Measuring error rates in genomic perturbation screens: gold standards for human functional genomics

Traver Hart¹, Kevin R Brown¹, Fabrice Sircoulomb², Robert Rottapel^{2,3,4} & Jason Moffat^{1,5,*}

291 core essential genes

927 non-essential genes

1000 non-targeting control sgRNAs

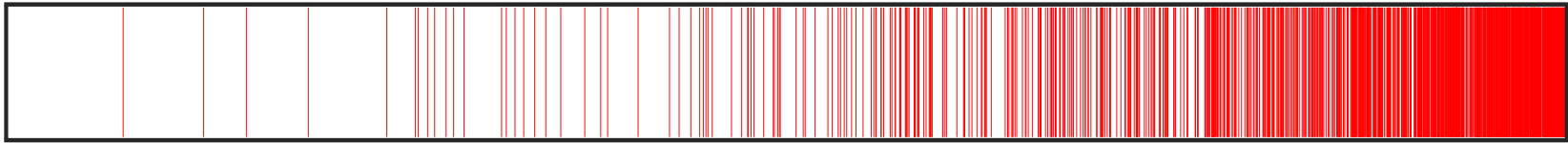
Are we doing any better? sgRNA sets to assess library performance



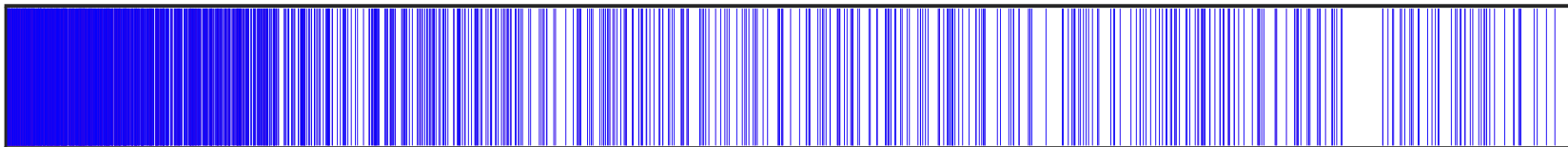
Most
depleted

← All sgRNAs, ranked →

Least
depleted

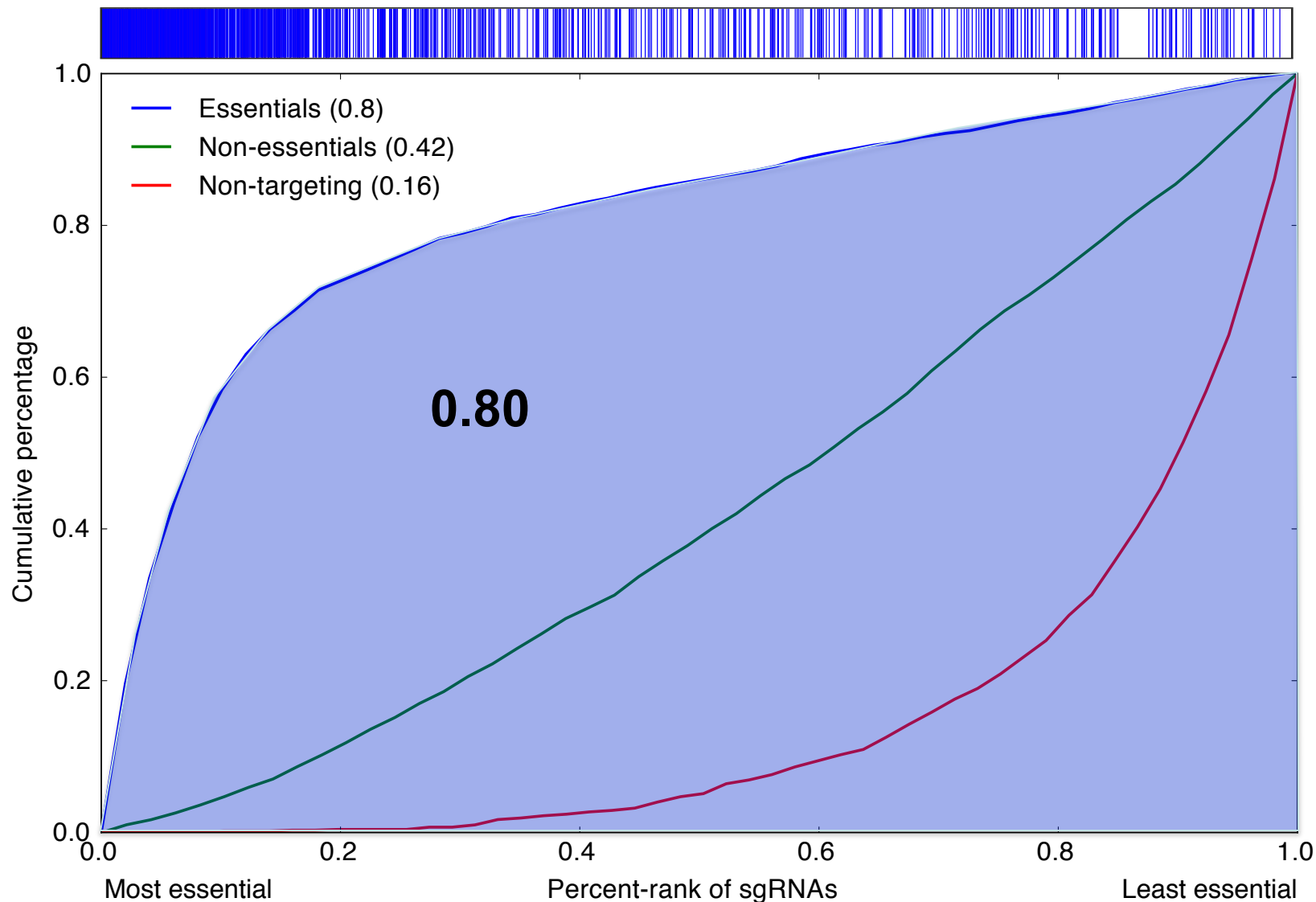


Non-targeting sgRNAs

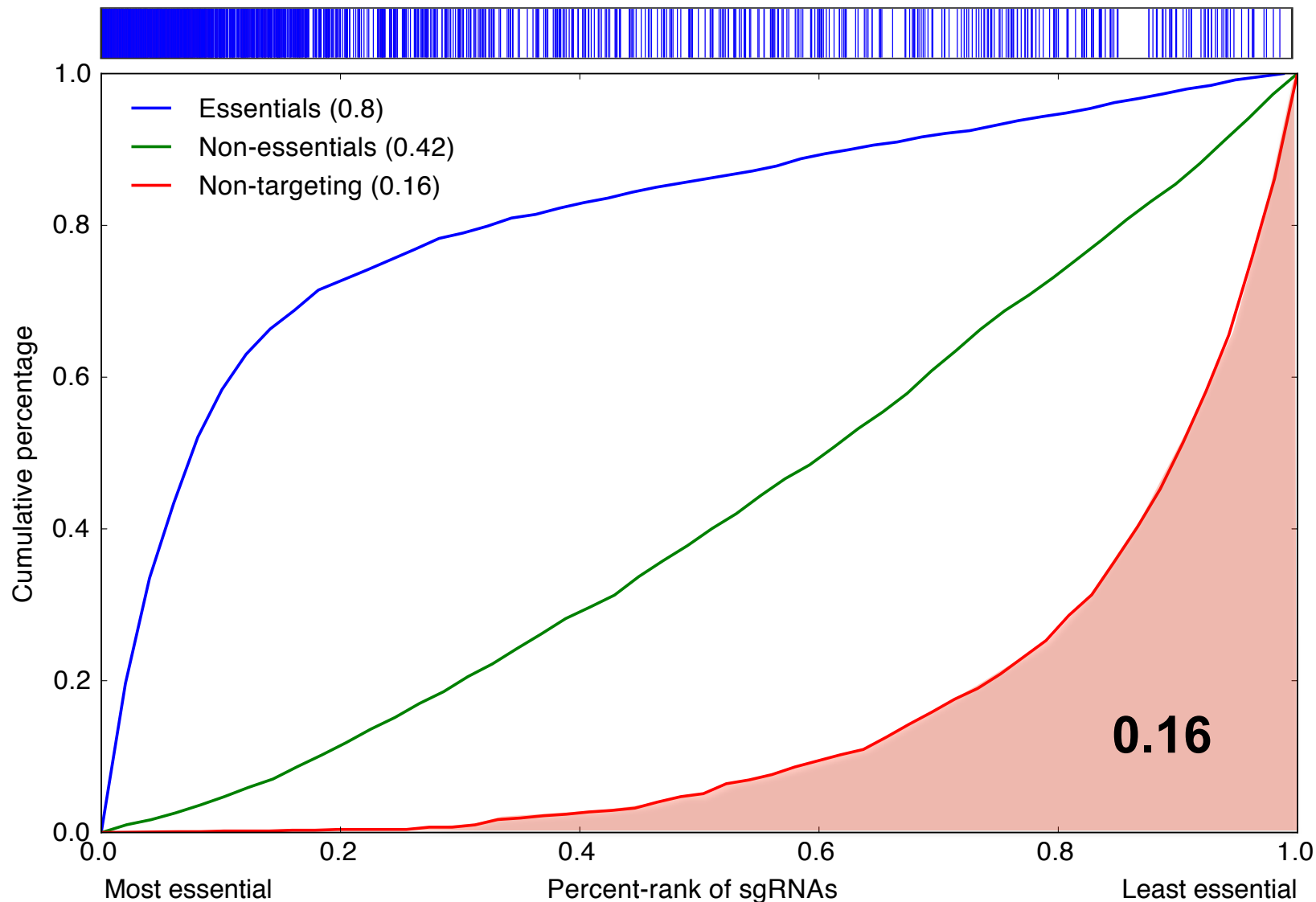


Core essential sgRNAs

Quantitate performance via Area Under Curve (AUC)



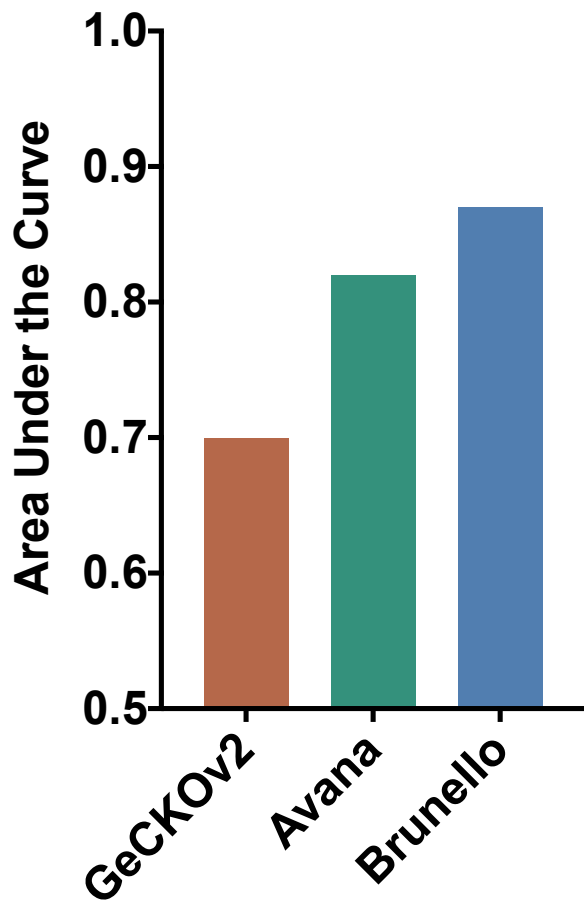
Quantitate performance via Area Under Curve (AUC)



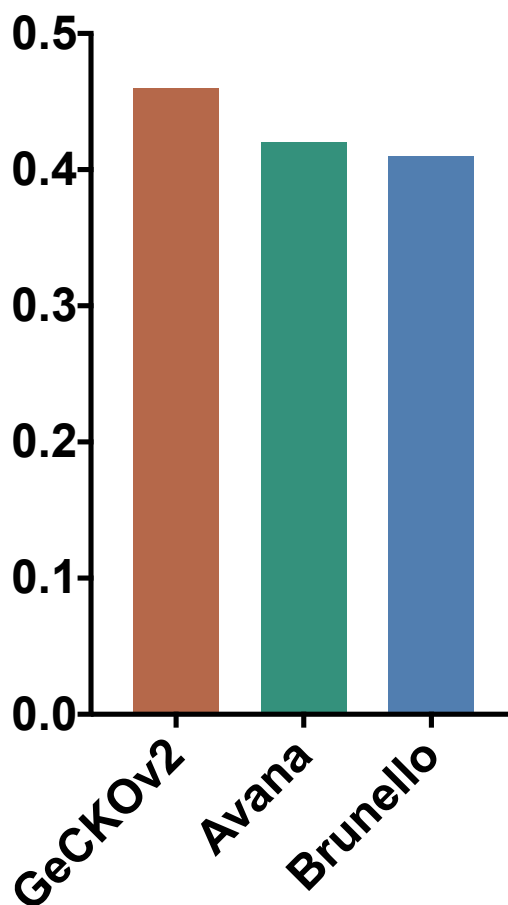
Improvement with each library



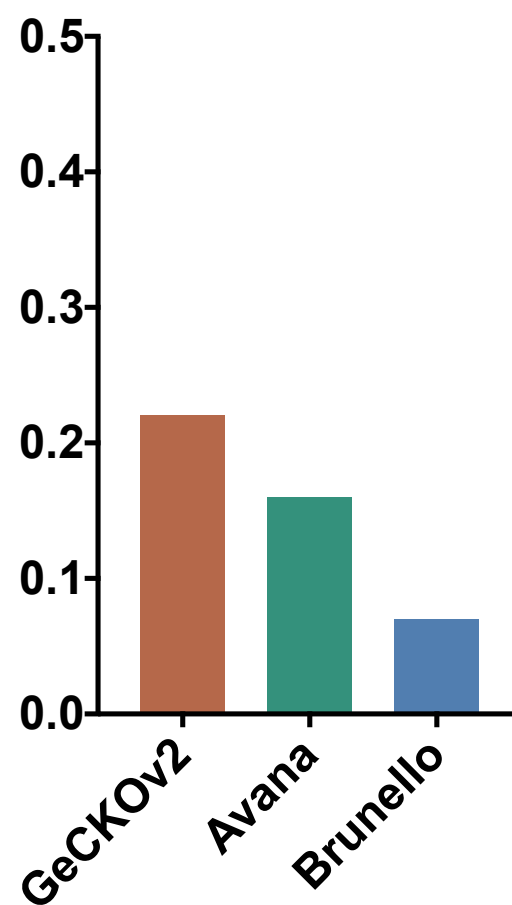
Essentials



Non-essentials

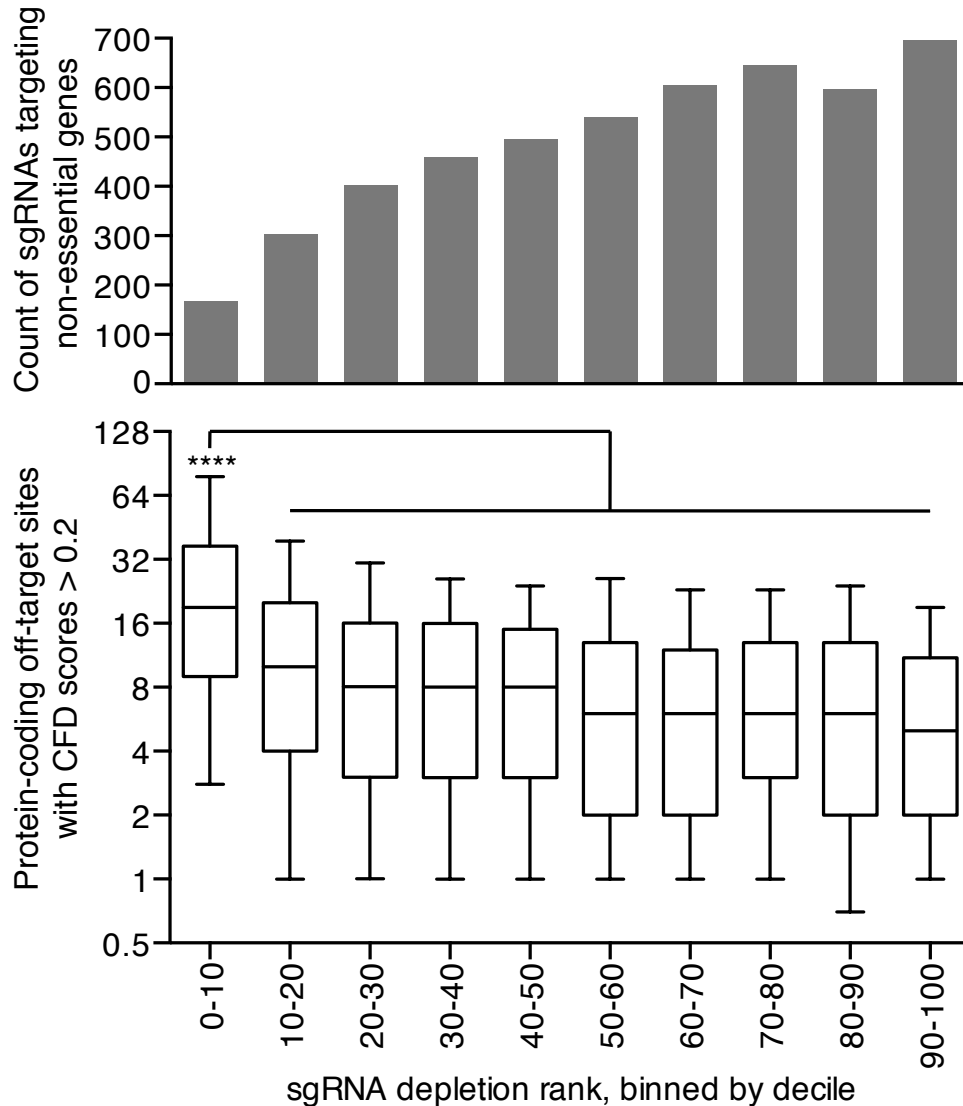


Non-targeting



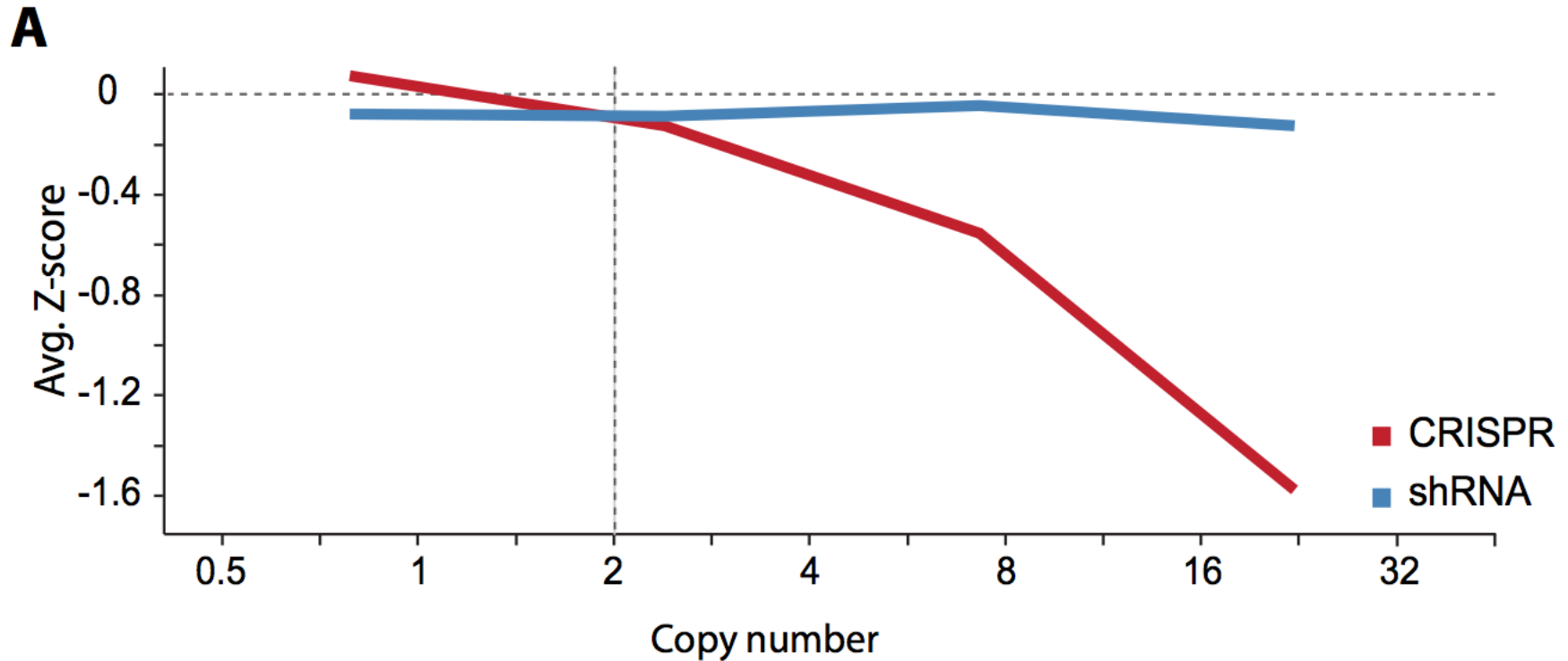
Promiscuous sgRNAs kill cells

Analysis of sgRNAs targeting non-essential genes

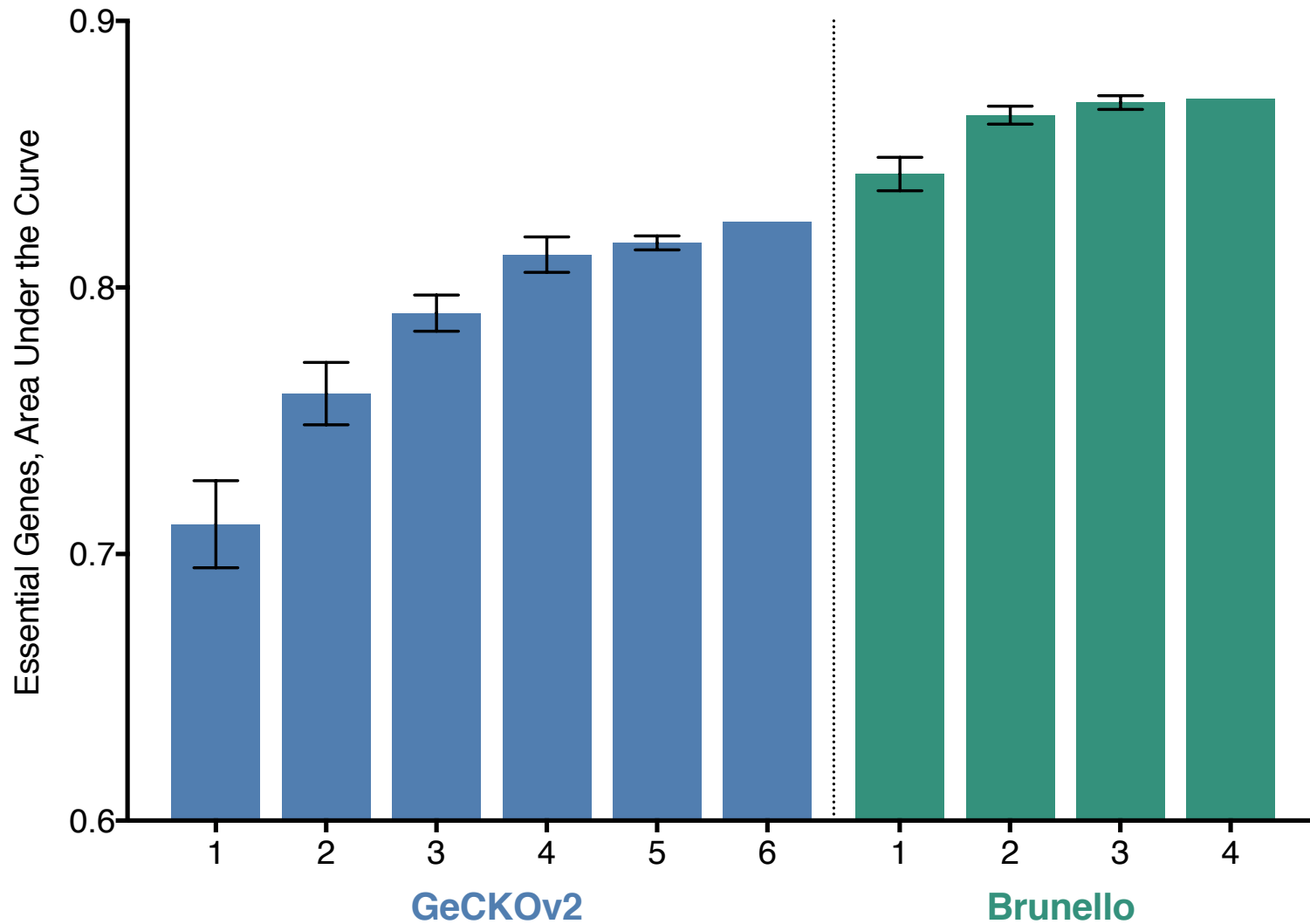


Doench, J. G. *et al.* Optimized sgRNA design to maximize activity and minimize off-target effects of CRISPR-Cas9. *Nat Biotechnol* **34**, 184–191 (2016).

Promiscuous sgRNAs kill cells... ... and copy number amplified regions



Subsampling to compare answers at the gene level: how many guides needed?



Number of sgRNAs per gene







Smaller, more active libraries



- Why smaller libraries? No doubt that more sgRNAs gives more data, better statistical significance. But it comes at a cost:
 - More cells: puts some models out of reach (e.g. primary cells, *in vivo* applications, flow cytometry)
 - More dollars: more cell culture, more sequencing, etc.
- Primary screen: identify genes that *might* be hits by casting a wide net for follow-up (low false negative rate)
- Secondary screen: more sgRNAs per gene result in a low false positive rate moving forward

What analysis *cannot* do for you

8. Matrix of sgRNA Abundance

sgRNA	Init.	Pos.	Neg.
	16	0	21
	19	100	19
	17	0	0
	18	0	26
	14	0	18
	16	0	16

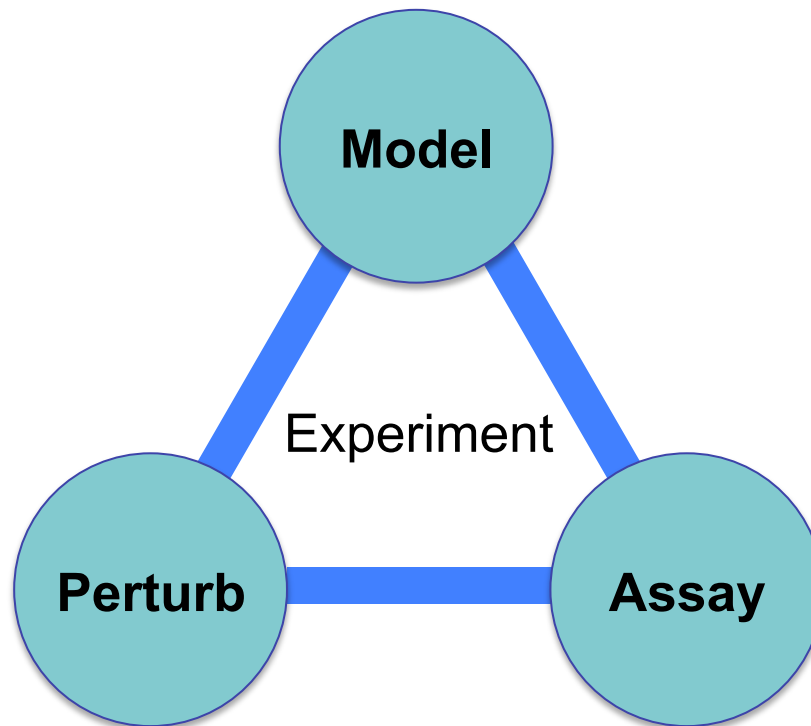
- Take technically poor data and make it technically sound
 - If your replicates don't line up well, “analysis” won't change that
- Tell you what assays to do next – you should have thought about this *before* the screen
 - Only rarely does the identity of genes in a hit list lead to an obvious hypothesis
- The #1 rule of screens with CRISPR: ***require multiple perturbations of independent sequence to give the same phenotype across all assays***

Functional Genomics

Pooled screens are great, you should do one!



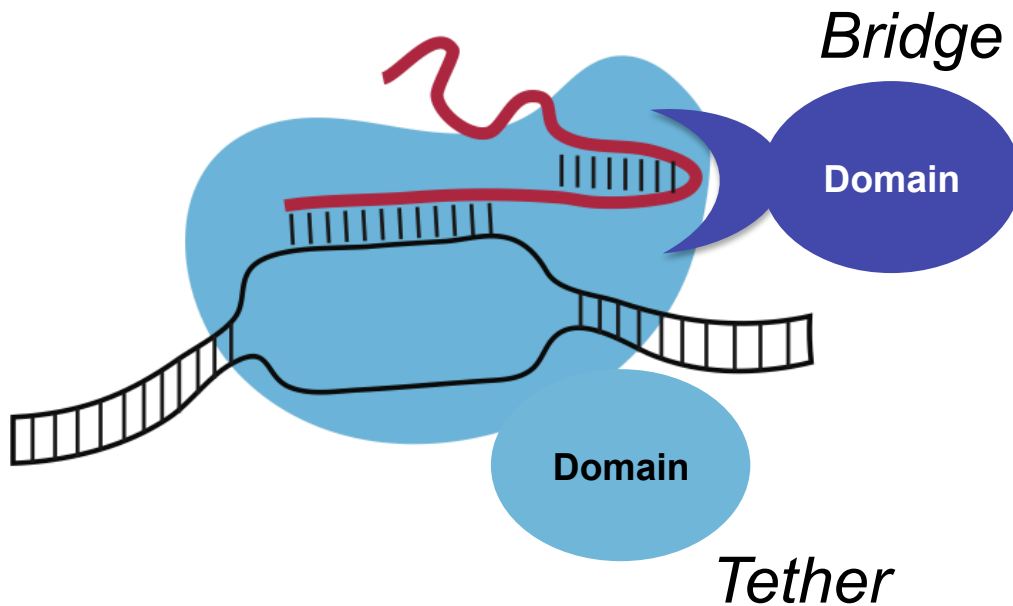
- CRISPR
(knockout)
- ORF
(overexpression)
- RNAi
(knockdown)
- CRISPRa
(overexpression)
- CRISPRi
(knockdown)



Diversity of Cas9 activities



Nuclease-dead Cas9 (dCas9)



Programmable DNA binding protein:

1. Tether protein domains to Cas9 itself
2. Bridge via an RNA stem loop engineered into tracrRNA along with fusion protein

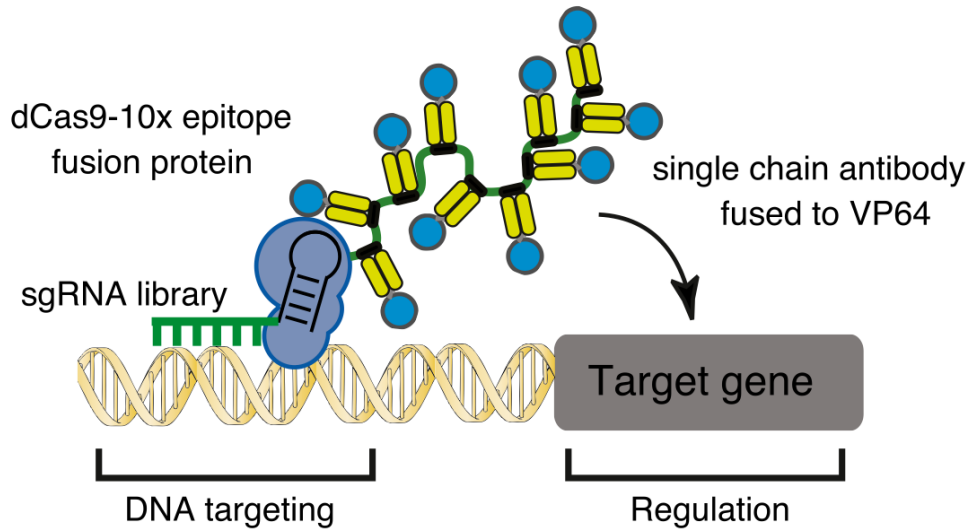
Activities

- Transcriptional activation (CRISPRa)
- Transcriptional repression (CRISPRi)
- Visualization
- Mutagenesis

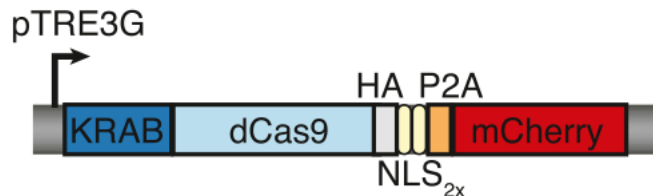
dCas9 for transcriptional modulation



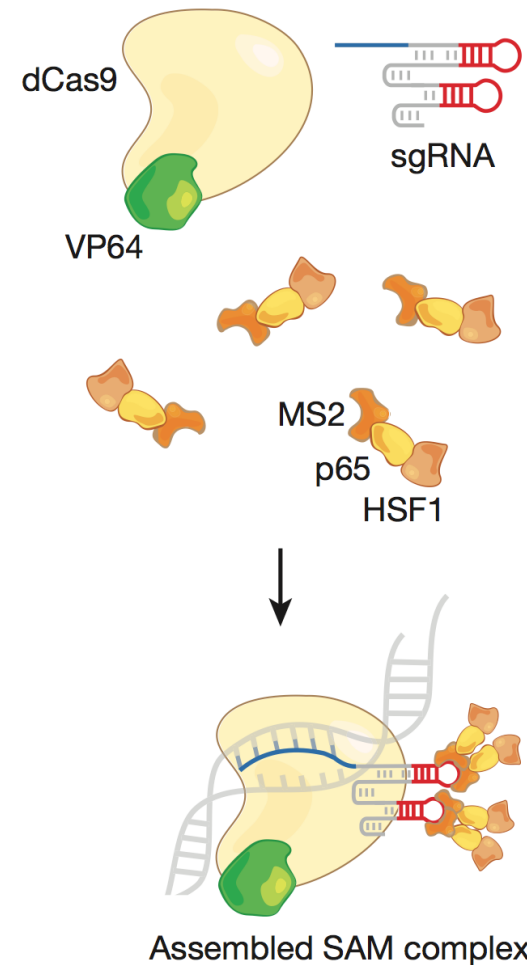
CRISPRa screening



CRISPRi

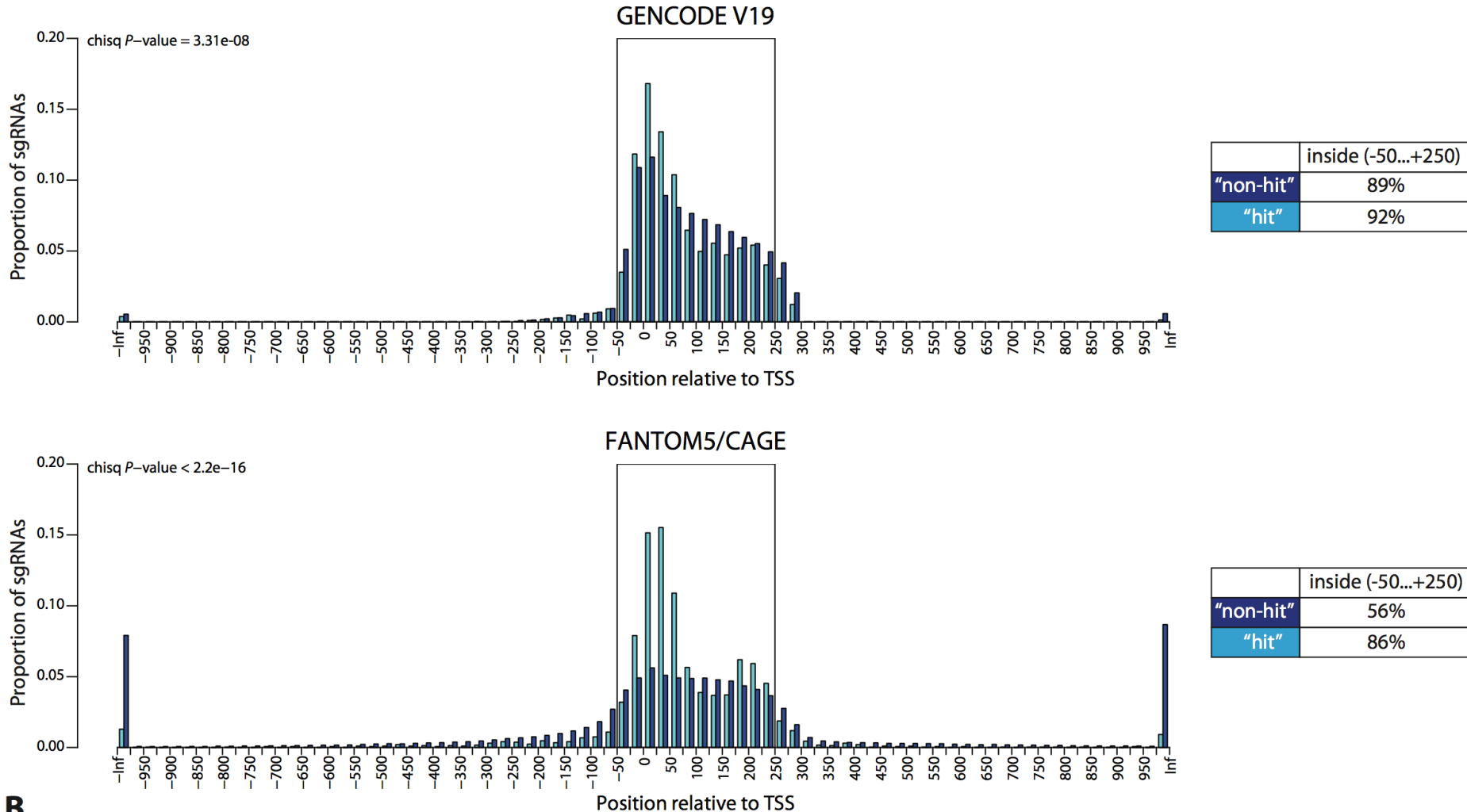


Gilbert et al., Cell, 2014



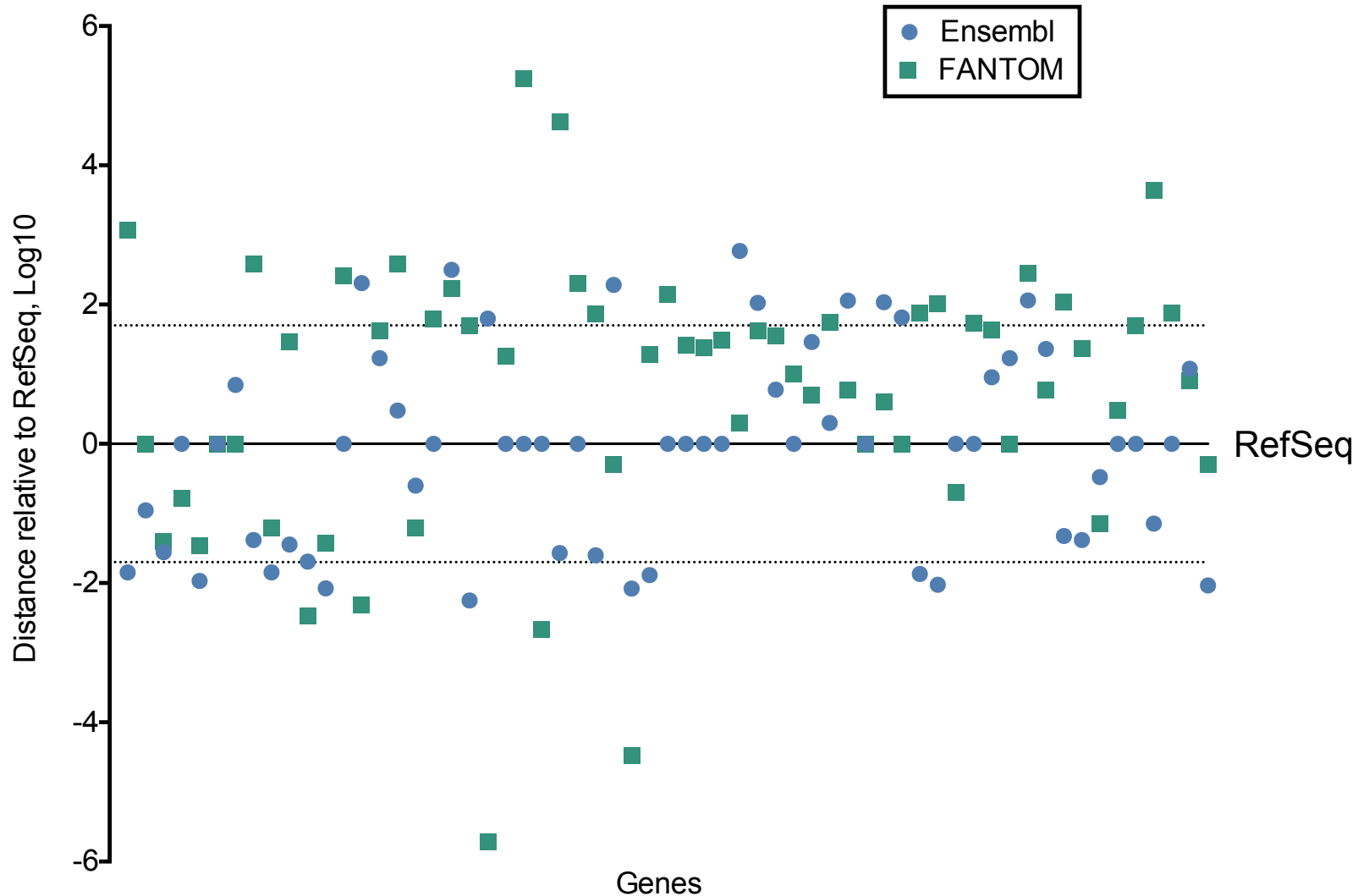
Konermann et al., Nature, 2014

CRISPRi (and CRISPRa) designs require good transcription start site (TSS) annotation

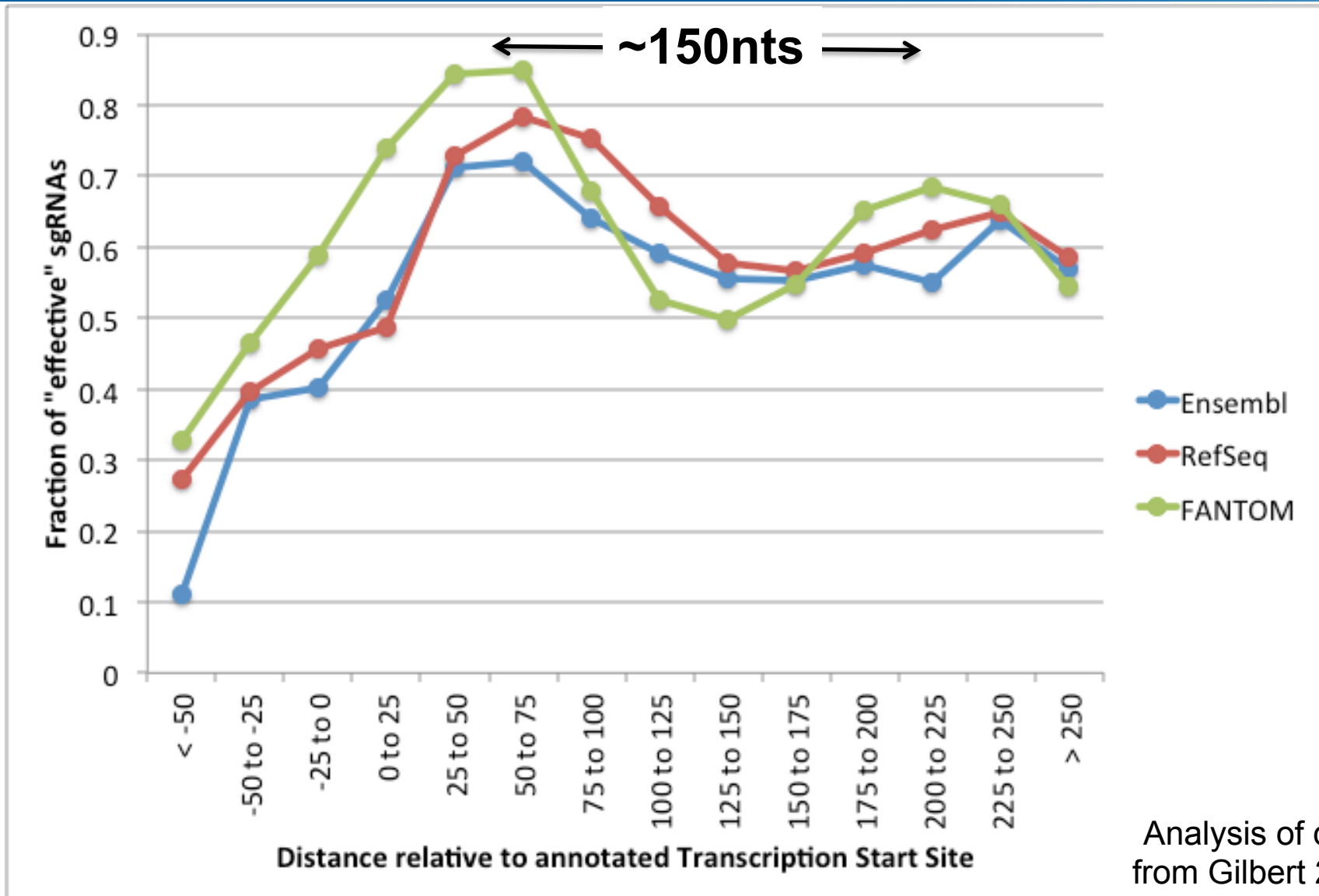


Radzisheuskaya, A., Shlyueva, D., Müller, I. & Helin, K. Optimizing sgRNA position markedly improves the efficiency of CRISPR/dCas9-mediated transcriptional repression. *Nucleic Acids Research* **44**, e141–e141 (2016).

Transcription Start Site (TSS) annotations differ across databases



CRISPRi location preference



CRISPR: what it is and what it isn't



- For genetic screens (i.e. functional genomics) CRISPR is a game-changer.
 - Much greater on-target efficacy
 - Many fewer off-target effects
- For gene *editing*, CRISPR makes it possible, but that is not the same thing as going from hard to easy.
 - Homology-directed repair low efficiency
 - Delivery challenges
 - Single cell cloning
- Like any tool or technology, need to understand its strengths and weaknesses when using it.

The Hype



“the capability to quickly and efficiently alter any gene sequence [will] have profound impacts on biological research and to yield potential therapeutic strategies for genetic diseases.” – *Nat. Rev.*

2013
TALENs

2003
RNAi

“within a few years [this technology] should yield a rough idea of what each of our genes does.” – *Fortune*

“a battle of biblical proportions over the patents for a crucial enzyme in molecular biology.” – *The Scientist*

1996
Taq polymerase

The Economist, 2015

Gartner, an American consultancy, has a simple yet elegant way of describing the life of a promising new technology. First, it is talked up to a **peak of inflated expectations**. Then it falls into a **trough of disillusionment**. After that, if it survives, it begins climbing the **slope of enlightenment**. Finally, it reaches the **plateau of productivity**. In the world of biotechnology CRISPR/Cas9 is still ascending towards peak expectations.

