



Controversy In Pharmacogenomics

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Summer Program in Biostatistics & Computational Biology



Outline

- Background and Introduction
- Analysis and Comparison of Results
- Takeaway/Conclusion
- Future Work
- Acknowledgements
- References



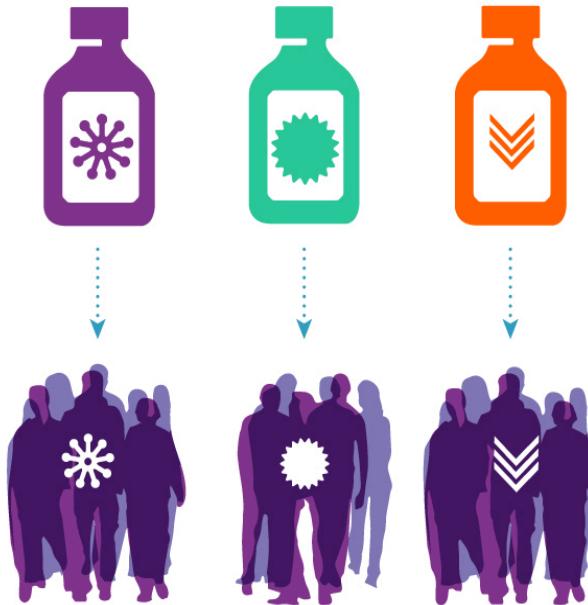
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“Doctors have always recognized that every patient is unique, and doctors have always tried to tailor their treatments as best they can to individuals. You can match a blood transfusion to a blood type — that was an important discovery. What if matching a cancer cure to our genetic code was just as easy, just as standard? What if figuring out the right dose of medicine was as simple as taking our temperature?”

- President Obama, January 30, 2015

UNDERSTANDING PRECISION MEDICINE

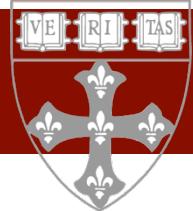
In precision medicine, patients with tumors that share the same genetic change receive the drug that targets that change, no matter the type of cancer.



Precision Medicine



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Great Insight... LETTER

doi:10.1038/nature11003

The Cancer Cell Line Encyclopedia enables predictive modelling of anticancer drug sensitivity

Jordi Barretina^{1,2,3†*}, Giordano Caponigro^{4*}, Nicolas Stransky^{1*}, Kavitha Venkatesan^{4*}, Adam A. Margolin^{1†*}, Sungjoon Kim⁵, Christopher J. Wilson⁴, Joseph Lehár⁴, Gregory V. Kryukov¹, Dmitriy Sonkin⁴, Anupama Reddy⁴, Manway Liu⁴, Lauren Murray¹, Michael F. Berger^{1†}, John E. Monahan⁴, Paula Morais¹, Jodi Meltzer⁴, Adam Korejwa¹, Judit Jané-Valbuena^{1,2}, Felipa A. Mapa⁴, Joseph Thibault⁵, Eva Bric-Furlong⁴, Pichai Raman⁴, Aaron Shipway⁵, Ingo H. Engels⁵, Jill Cheng⁶, Guoying K. Yu⁶, Jianjun Yu⁶, Peter Aspesi Jr⁴, Melanie de Silva⁴, Kalpana Jagtap⁴, Michael D. Jones⁴, Li Wang⁴, Charles Hatton³, Emanuele Palescandolo³, Supriya Gupta¹, Scott Mahan¹, Carrie Sougnez¹, Robert C. Onofrio¹, Ted Liefeld¹, Laura MacConaill³, Wendy Winckler¹, Michael Reich¹, Nanxin Li⁵, Jill P. Mesirov¹, Stacy Barbara L. Weber⁴, Jeff Porter⁴, Markus Warmuth¹, Michael P. Morrissey^{4*}, William R. Sellers^{4*}, Ronald A. DePinho¹

ARTICLE

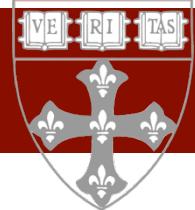
doi:10.1038/nature11005

Systematic identification of genomic markers of drug sensitivity in cancer cells

Mathew J. Garnett^{1*}, Elena J. Edelman^{2*}, Sonja J. Heidorn^{1*}, Chris D. Greenman^{1†}, Anahita Dastur², King Wai Lau¹, Patricia Greninger², I. Richard Thompson¹, Xi Luo², Jorge Soares¹, Qingsong Liu^{3,4}, Francesco Iorio^{1,5}, Didier Surdez⁶, Li Chen², Randy J. Milano⁶, Graham R. Bignell¹, Ah T. Tam², Helen Davies¹, Jesse A. Stevenson², Syd Barthorpe¹, Stephen R. Lutz², Fiona Kogera¹, Karl Lawrence¹, Anne McLaren-Douglas¹, Xeni Mitropoulos², Tatiana Mironenko¹, Helen Thi², Laura Richardson¹, Wenjun Zhou^{3,4}, Frances Jewitt¹, Tinghu Zhang^{3,4}, Patrick O'Brien¹, Jessica L. Boisvert², Stacey Price¹, Wooyoung Hur^{3,4}, Wanjuan Yang¹, Xianming Deng^{3,4}, Adam Butler¹, Hwan Geun Choi^{3,4}, Jae Won Chang^{3,4}, Jose Baselga², Ivan Stamenkovic⁷, Jeffrey A. Engelmann², Sreenath V. Sharma^{2†}, Olivier Delattre⁶, Julio Saez-Rodriguez⁵, Nathanael S. Gray^{3,4}, Jeffrey Settleman², P. Andrew Futreal¹, Daniel A. Haber^{2,8}, Michael R. Stratton¹, Sridhar Ramaswamy², Ultan McDermott¹ & Cyril H. Benes²



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... Only one problem

ANALYSIS

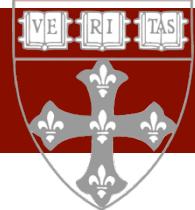
doi:10.1038/nature12831

Inconsistency in large pharmacogenomic studies

Benjamin Haibe-Kains^{1,2}, Nehme El-Hachem¹, Nicolai Juul Birkbak³, Andrew C. Jin⁴, Andrew H. Beck^{4*}, Hugo J. W. L. Aerts^{5,6,7*}
& John Quackenbush^{5,8*}



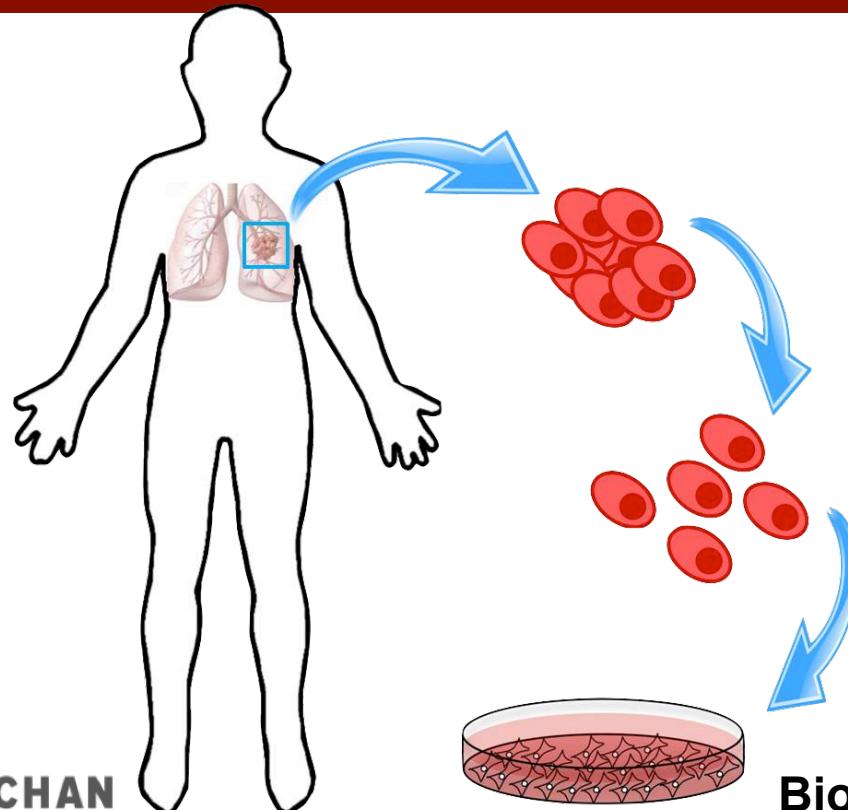
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Obtaining Samples



Cell Lines

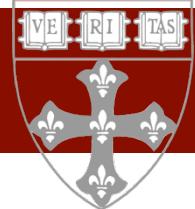


- Retrieve and genetically transform cell cultures derived from various tissue
- Expose samples to drugs
- Record Drug Response (sensitivity)
- Quantify the effectiveness of the drug (does it work?)

Biological Replicates

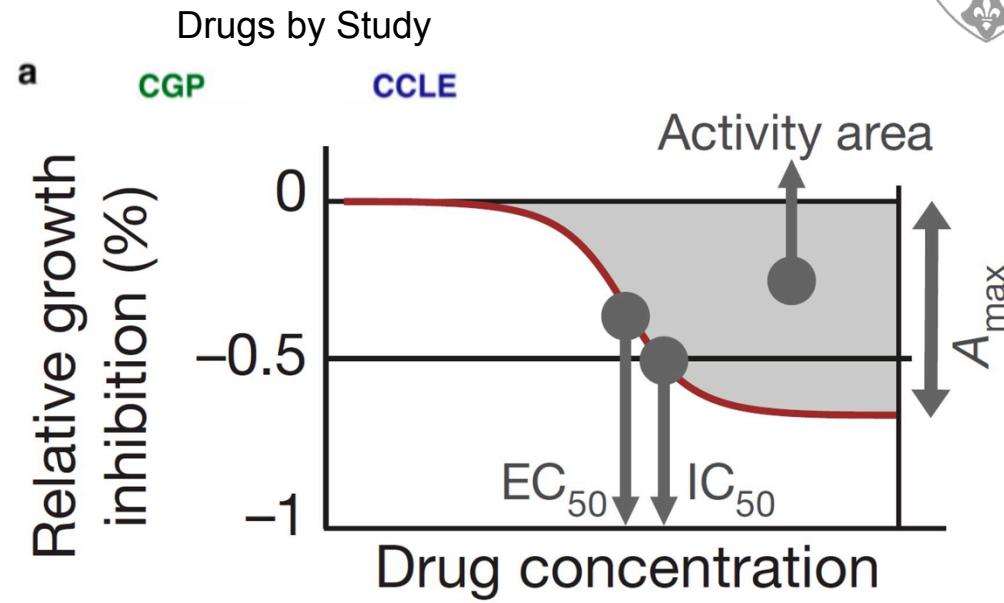


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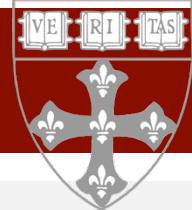
How They Did It...

- 471 cell lines shared between both data sets for 64 gene mutations
- Compared drug sensitivity between CGP and CCLE for 15 drugs
- IC_{50} : Drug concentration necessary to inhibit 50% of growth



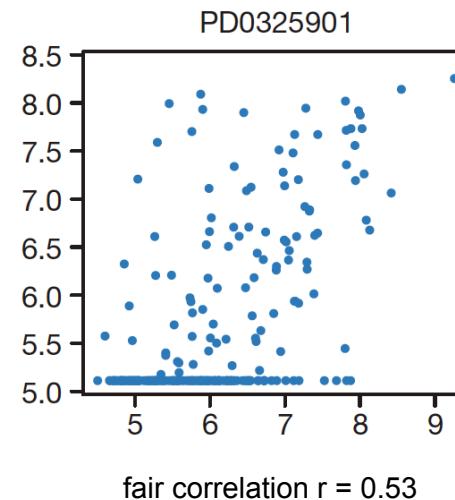
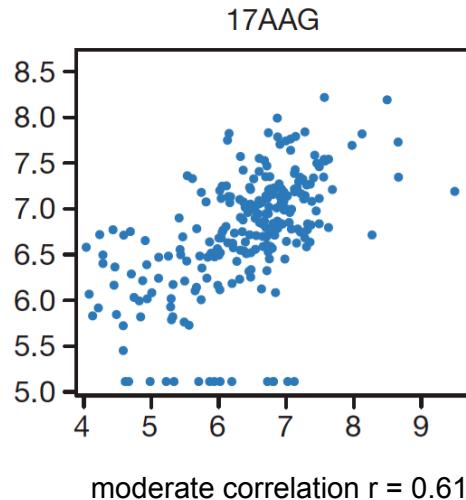
Barretina, J. et al. (2012) Kains et al. (2013)

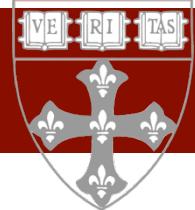




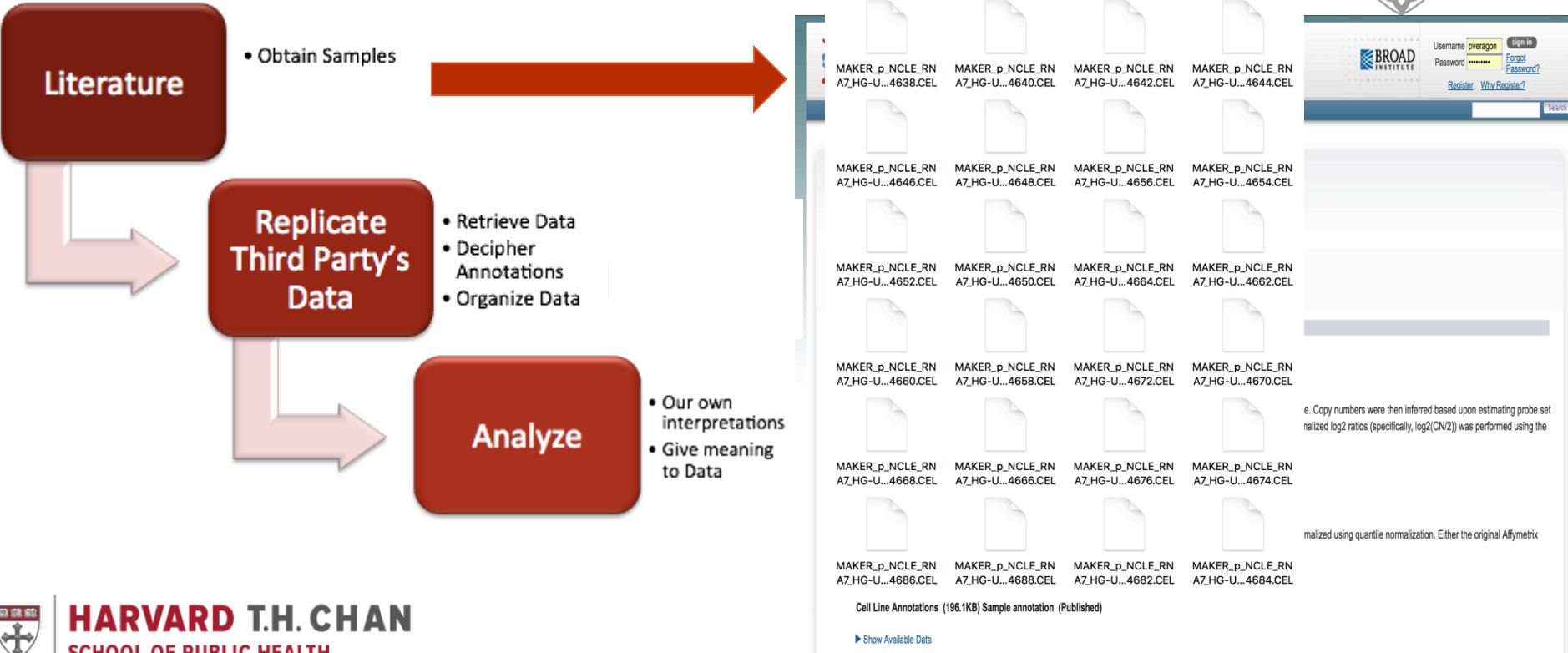
Close But not Quite...

- Ranked the response of the 471 shared cell lines and computed the Spearman's rank correlations for the IC₅₀ values.





Our Interpretation of The Problem



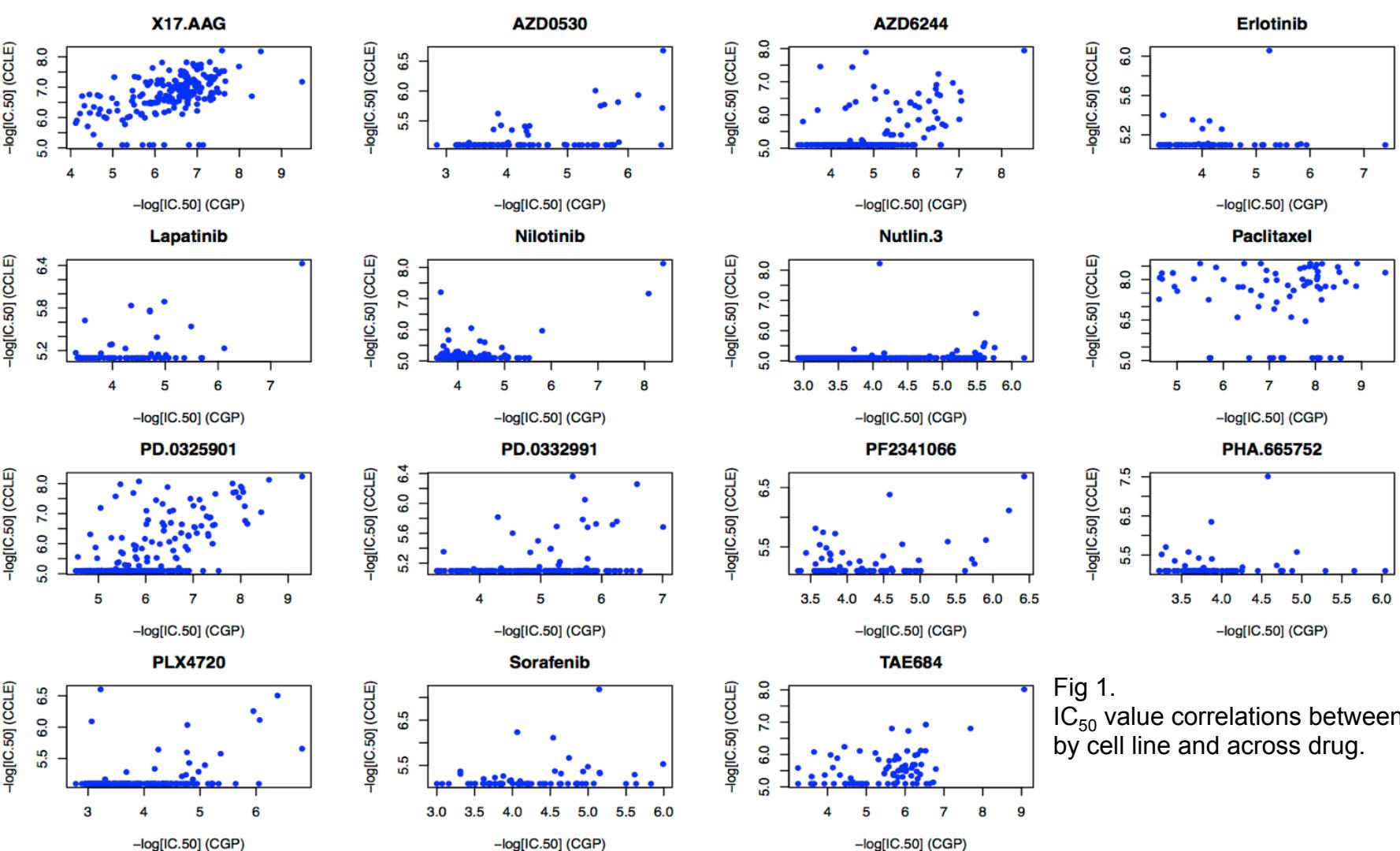


Fig 1.
 IC_{50} value correlations between datasets by cell line and across drug.

Haibe-Kains et al. (2013) -- Left
Our results -- Right

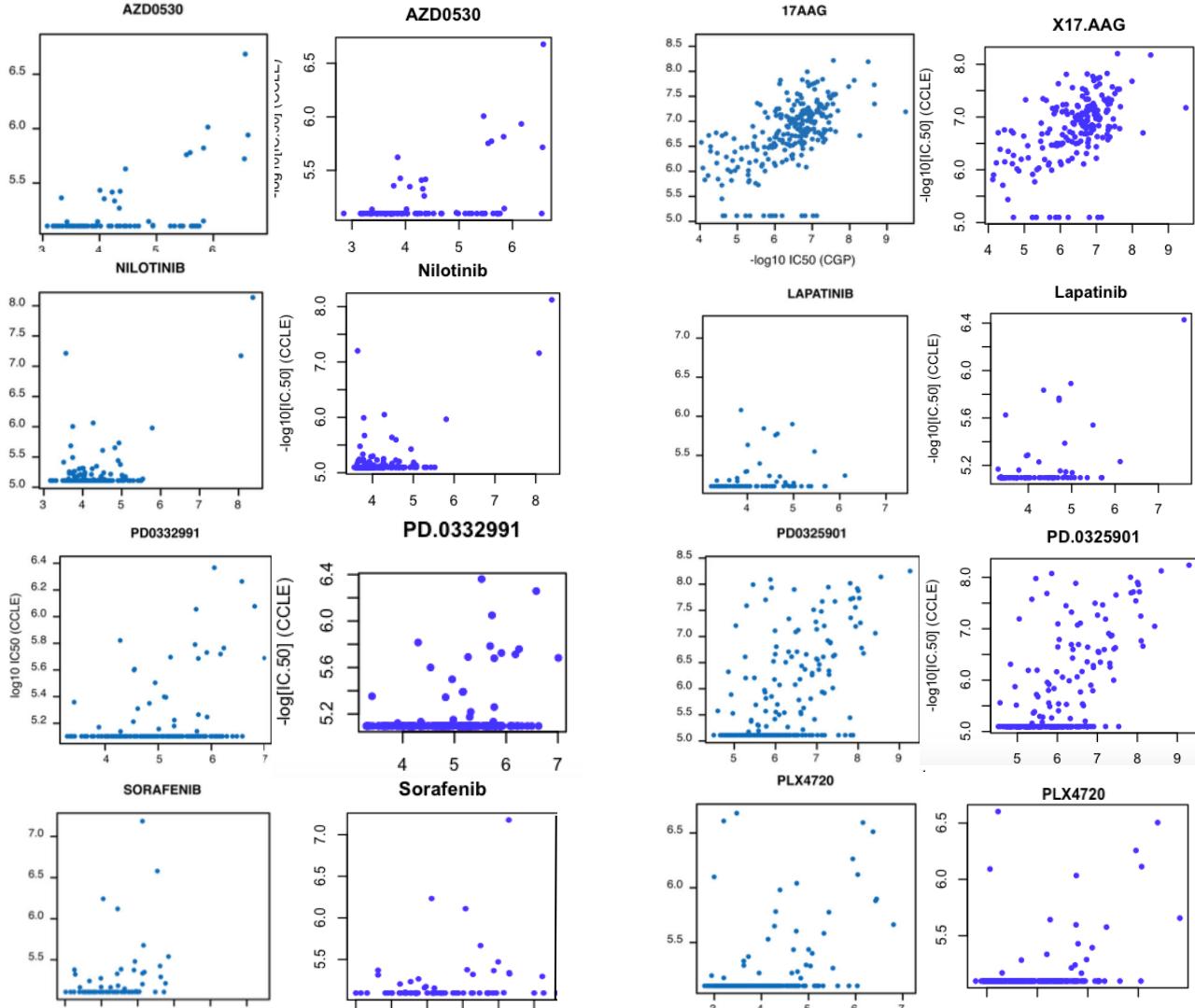
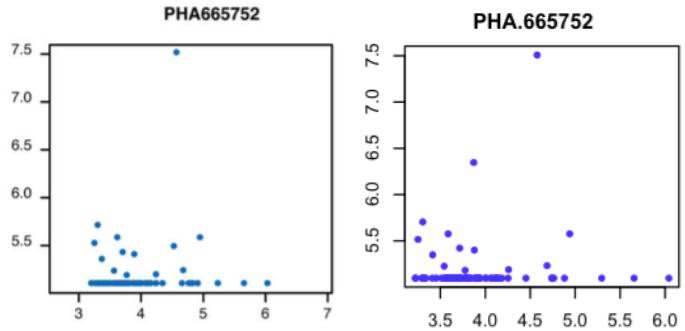
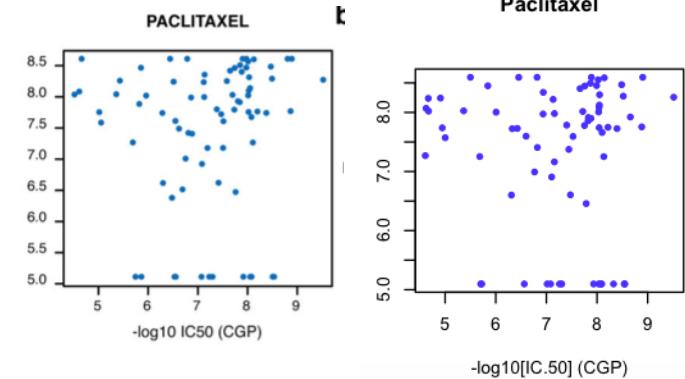
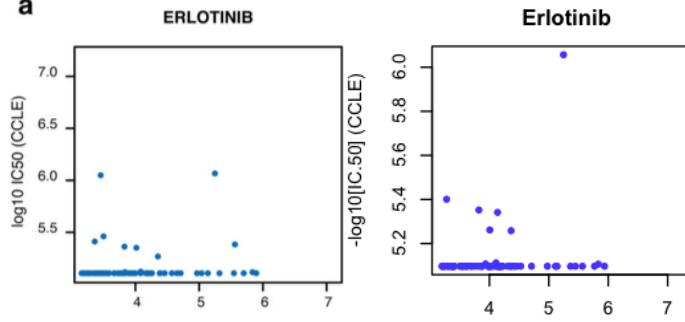
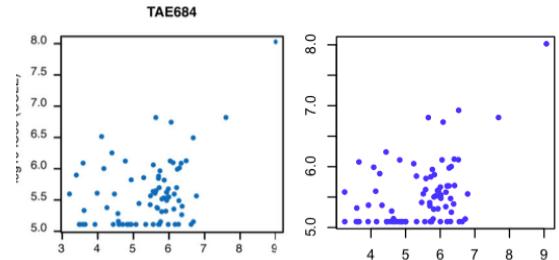
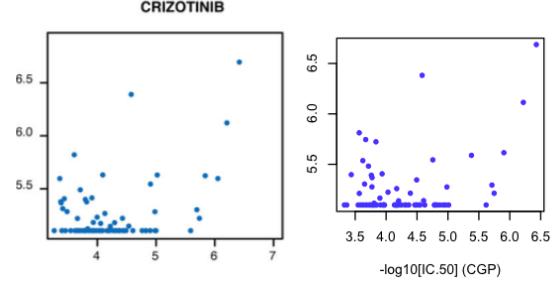
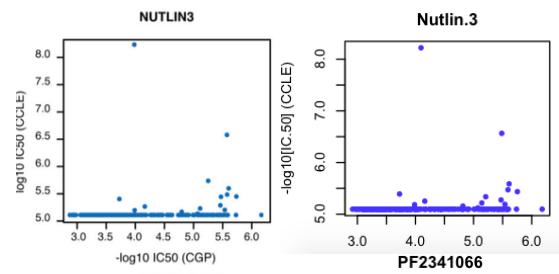
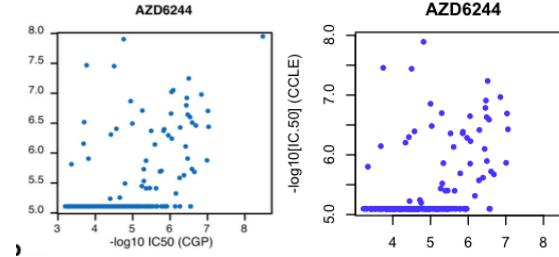


Fig 2.
IC₅₀ value comparisons
between our analysis and
Haibe-Kains et al. (2013).





Comparison between Spearman Correlation Coefficients

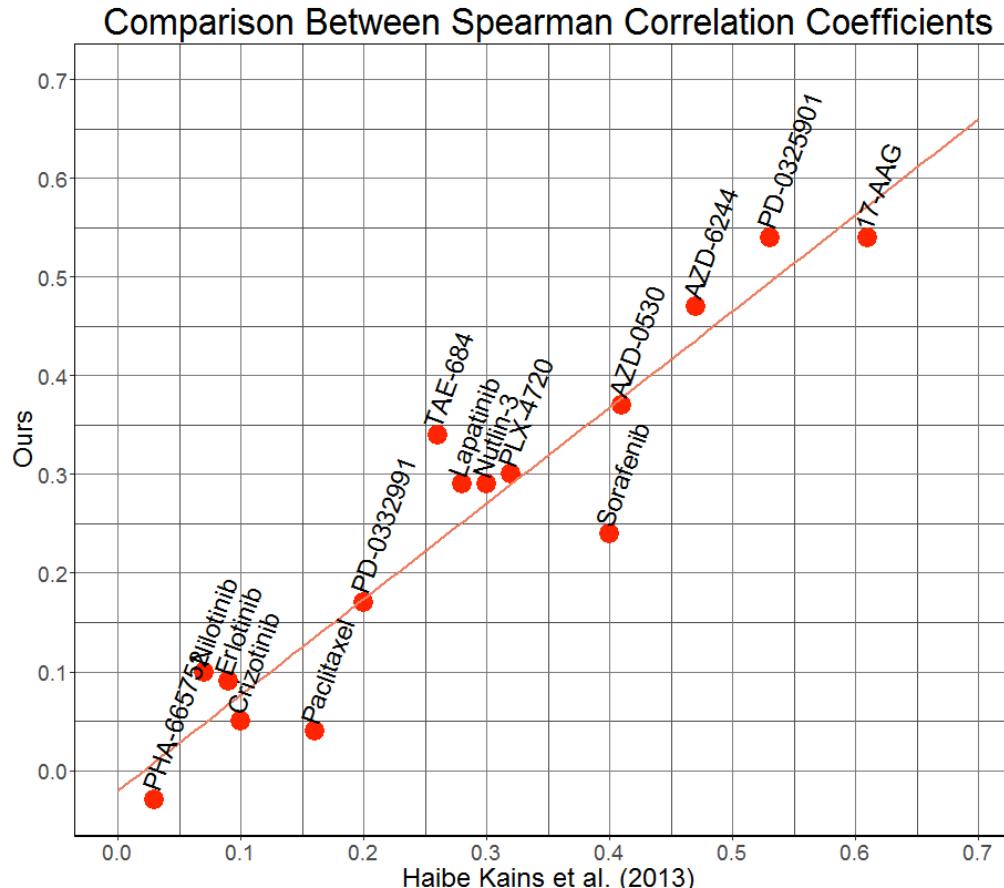
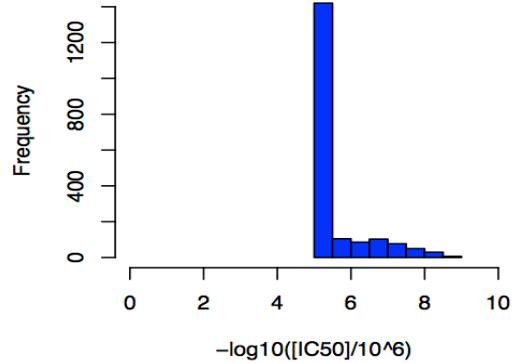


Fig 3. Scatterplot of drug IC₅₀ value correlations comparing ours and the literature's.



Transformed CCLE IC50 Values



CGP IC50 Values

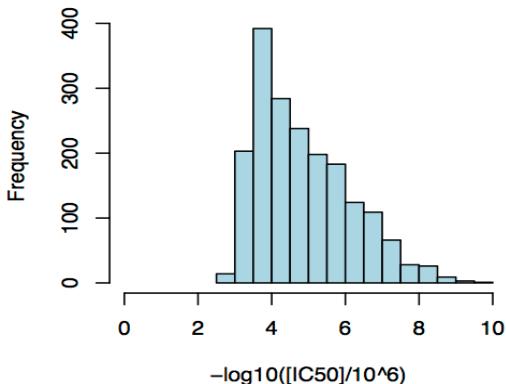
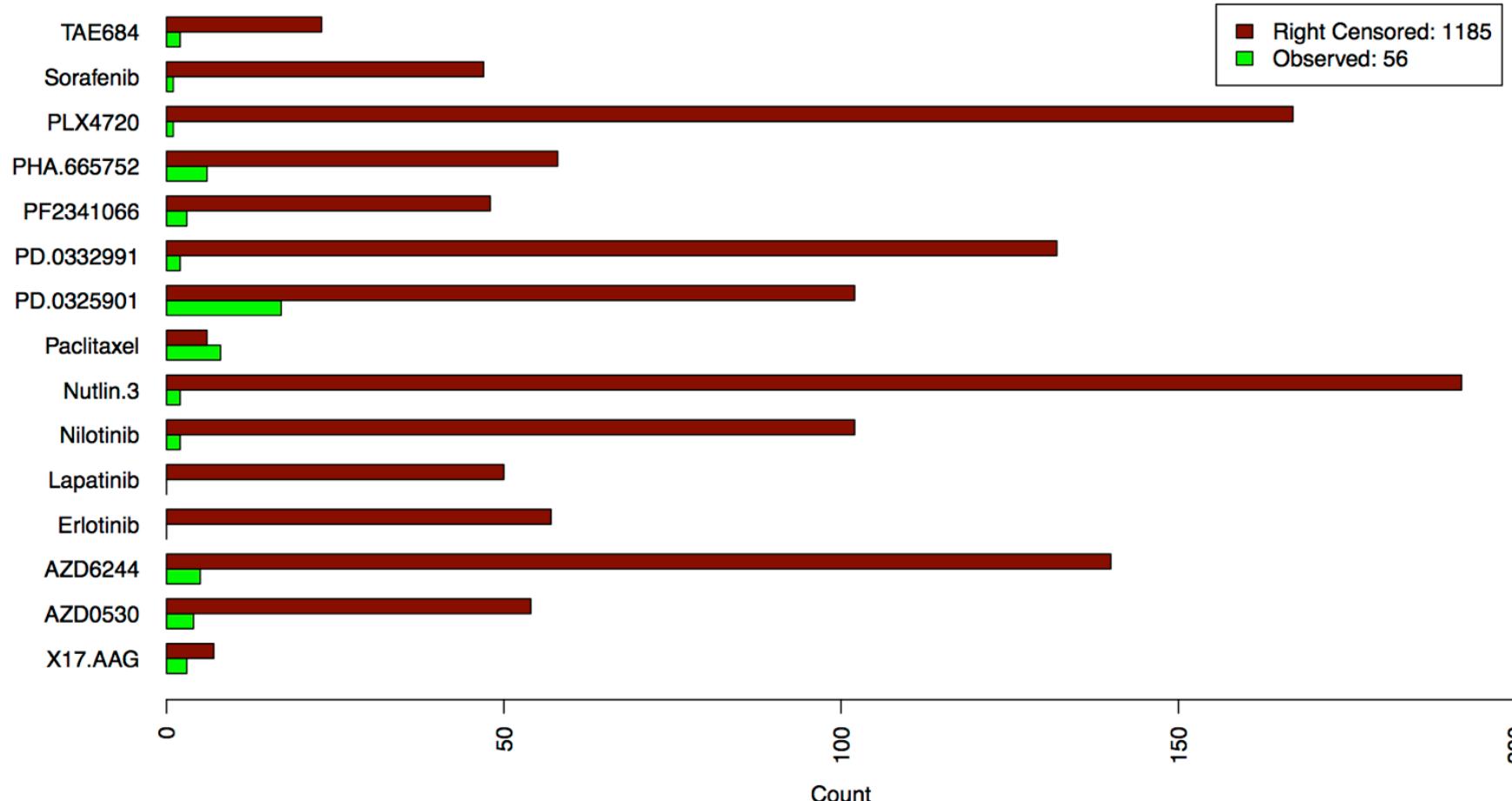


Fig 4.
 IC_{50} value distributions
by dataset.



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Observed vs. Right Censored Eights



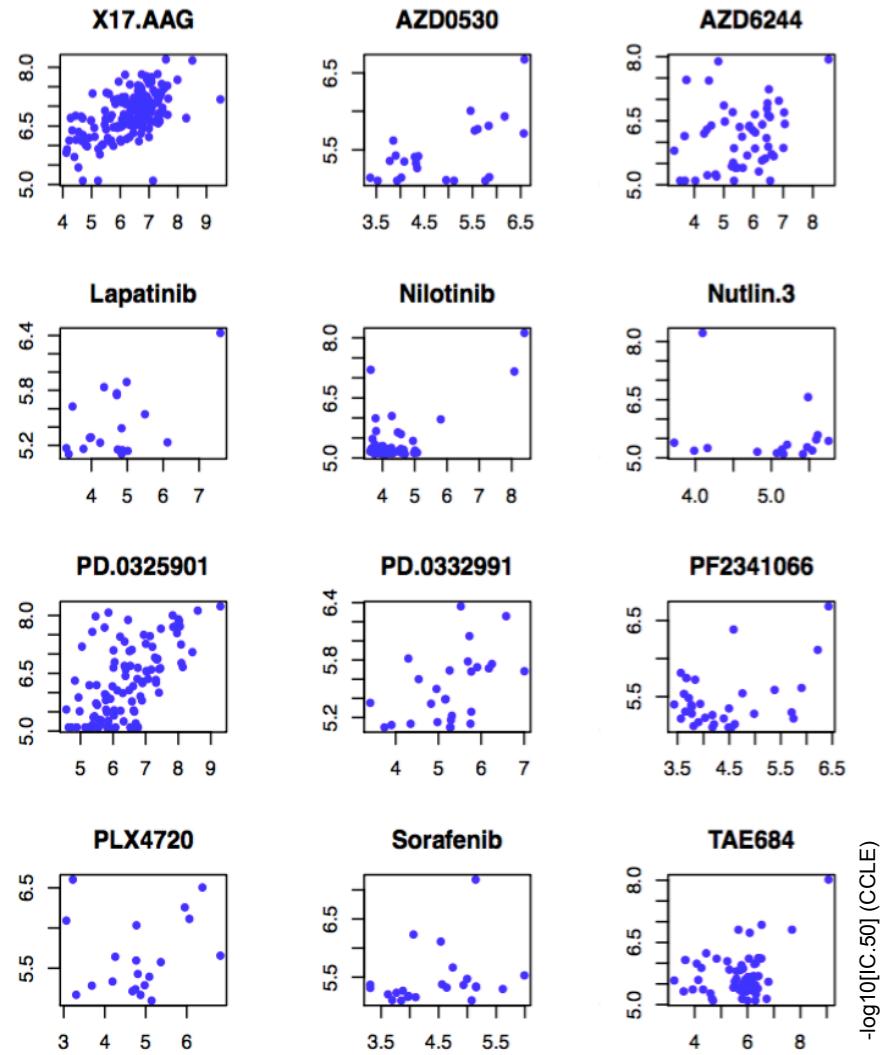
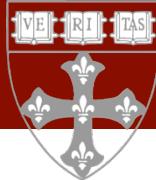


Table 1: Adjusted Correlations

	Pearson Adjusted	Spearman Adjusted
X17.AAG	.57	.55
AZD0530	.61	.48
AZD6244	.26	.27
Erlotinib	.41	.03
Lapatinib	.22	-.28
Nilotinib	.54	.19
Nutlin.3	.68	-.01
Paclitaxel	-.23	.26
PD.0325901	-.08	.02
PD.0332991	.58	.55
PF2341066	.48	.46
PHA.665752	.06	-.10
PLX4720	.15	.18
Sorafenib	.27	.39
TAE684	.37	.14

Fig 5.
Adjusted IC_{50} value correlations by cell line
and across drug.



Correlation table across studies

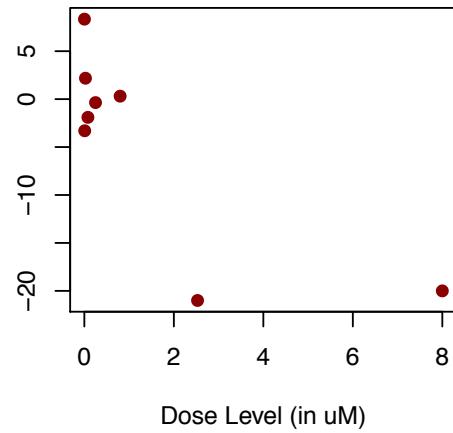
Table 2: Correlations Across Study and by Drug

	Pearson	Pearson Adj	Spearman	Spearman Adj	Spearman Haibe-Kains (2013)
X17.AAG	.51	.57	.54	.55	.61
AZD0530	.48	.61	.37	.48	.41
AZD6244	.49	.26	.47	.27	.47
Erlotinib	.37	.41	.05	.03	.10
Lapatinib	.11	.22	.09	-.28	.09
Nilotinib	.48	.54	.29	.19	.28
Nutlin.3	.58	.68	.10	-.01	.07
Paclitaxel	.12	-.23	.29	.26	.30
PD.0325901	-.05	-.08	.04	.02	.16
PD.0332991	.63	.58	.54	.55	.53
PF2341066	.26	.48	.17	.46	.20
PHA.665752	.09	.06	-.03	-.10	.03
PLX4720	.36	.15	.30	.18	.32
Sorafenib	.24	.27	.24	.39	.40
TAE684	.45	.37	.34	.14	.26

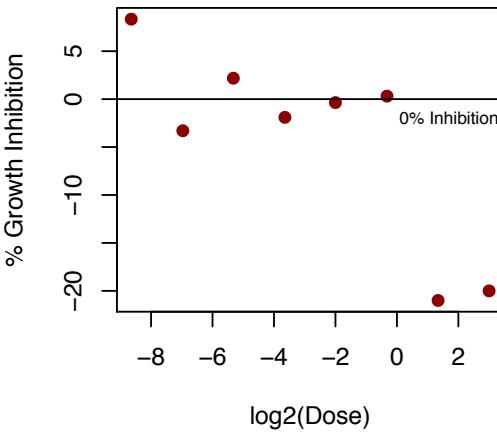
*Adj = Adjusted



Cell Line Raw Data

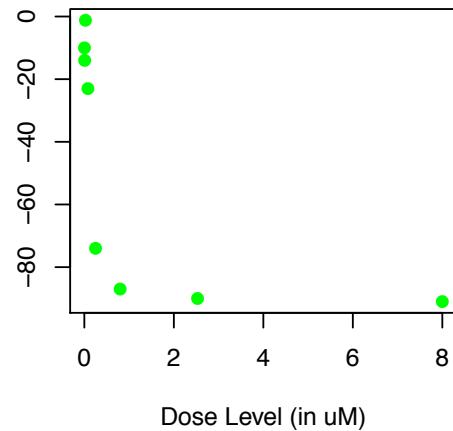


Drug Ineffective

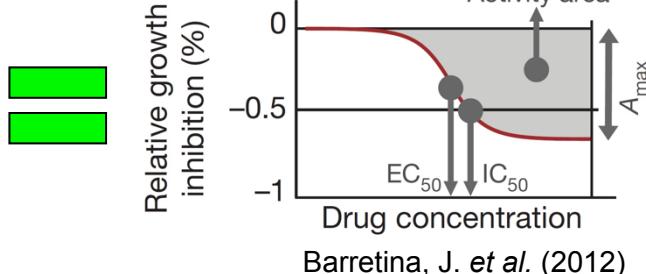
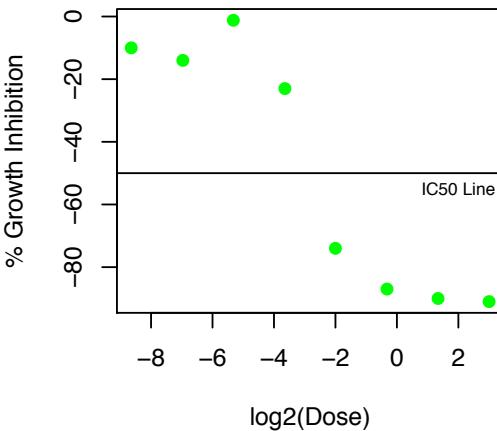


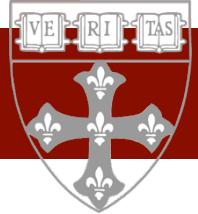
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Cell Line Raw Data

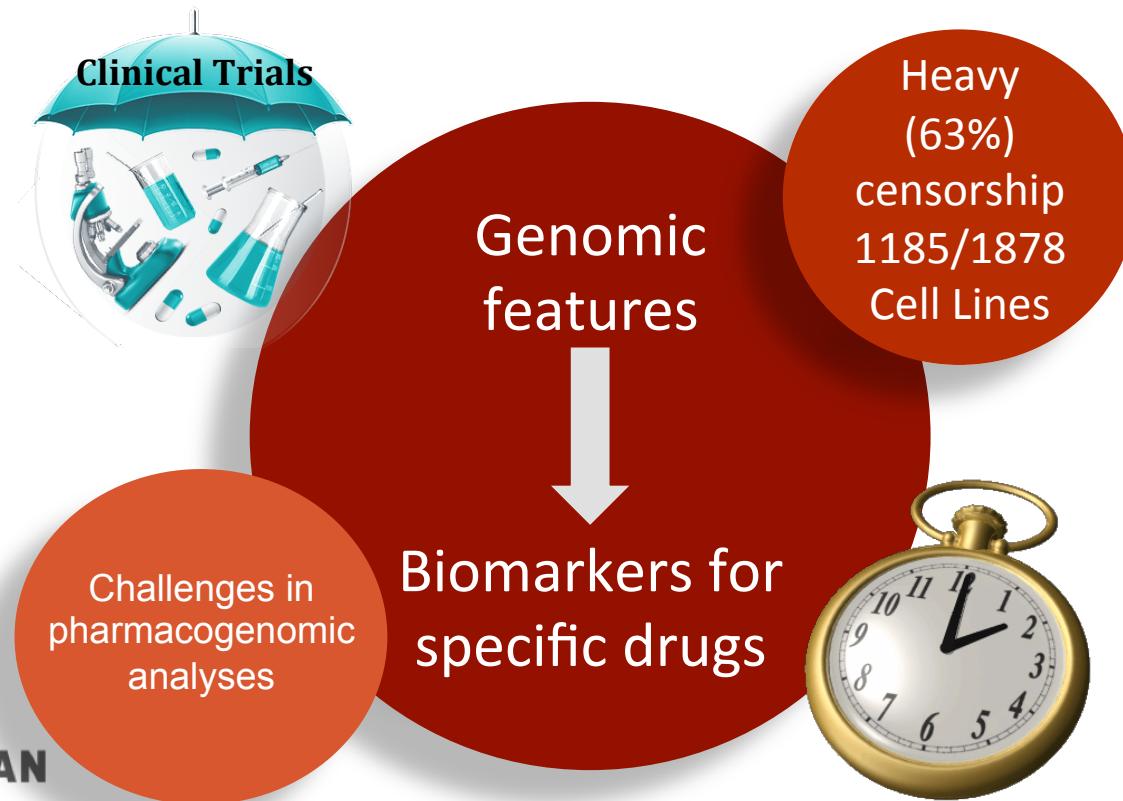


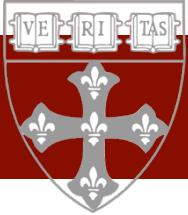
Drug Effective





Takeaway/Conclusion





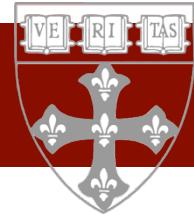
Future Work

Obtain gene expressions of both studies from raw .CEL files

Compare genomic features between studies across all cell lines

Better statistical summary of the curves for drug responses





Acknowledgments

- Dana-Farber Cancer Institute
- Rafael Irizarry, PhD
- William Townes, MS
- Summer Program in Biostatistics and Computational Biology
- Jessica Boyle, Heather Mattie, Olivia Orta
Dr. Rebecca Betensky
- Harvard T.H. Chan School of Public Health

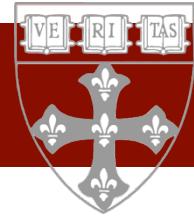


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- Garnett, Mathew J., et al. "Systematic identification of genomic markers of drug sensitivity in cancer cells." *Nature* 483.7391 (2012): 570-575.
- Haibe-Kains, Benjamin, et al. "Inconsistency in large pharmacogenomic studies." *Nature* 504.7480 (2013): 389-393.





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Dichotomization of IC₅₀ Values by Drug

Table 3 : Dichotomization of IC₅₀ Values by Drug

	Count: ++	Count: - +	Count: --	Count:+ -	(χ^2_1 , pval)	Fisher (95% CI)
X17.AAG	65	29	65	29	(26.0, <.001)	(2.6, 9.8)
AZD0530	14	5	34	24	(4.8, <.05)	(1.1, 15.7)
AZD6244	39	8	88	57	(25.4, <.001)	(3.1, 19.8)
Erlotinib	14	15	25	26	(0.0, NS)	(0.3, 2.5)
Lapatinib	6	3	30	27	(0.5, NS)	(0.4, 14.9)
Nilotinib	13	6	29	21	(2.9, NS)	(0.9, 11.1)
Nutlin.3	24	19	55	49	(0.6, NS)	(0.7, 3.1)
Paclitaxel	13	1	103	91	(9.3, <.01)	(2.1, 629.4)
PD.0325901	21	15	21	15	(1.4, NS)	(0.7, 5.6)
PD.0332991	62	25	78	41	(25.8, <.001)	(2.5, 9.0)
PF2341066	16	9	71	63	(1.8, NS)	(0.8, 5.5)
PHA.665752	6	7	32	32	(0.0, NS)	(0.2, 3.4)
PLX4720	15	4	90	78	(6.0, <.05)	(1.3, 18.5)
Sorafenib	13	9	26	22	(0.6, NS)	(0.6, 5.4)
TAE684	26	14	26	14	(6.1, <.05)	(1.3, 9.6)