Identification of Genes that can Selectively Kill Cancer Cells Using Boolean Implications Yang Gao¹, Subarna Sinha², David L. Dill²

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Motivation

- Many cancer treatment drugs do not target only cancer cells, which makes treatment less effective as well as generates serious side effects.
- Identifying drugs that only target cancer cells requires expensive and time-consuming experiments.
- Computational techniques can immensely aid in the process.
- We predict candidate genes whose altered expression can be lethal to cancer cells with a particular genetic alteration.

Background

- Boolean Implications[1] capture asymmetrical relations between pairs.
- Statistical test is used to test sparseness of each quadrant.
- Synthetic lethality (SL) arises when a combination of mutations in two genes leads to cell death, whereas a mutation in only one of these genes does not.

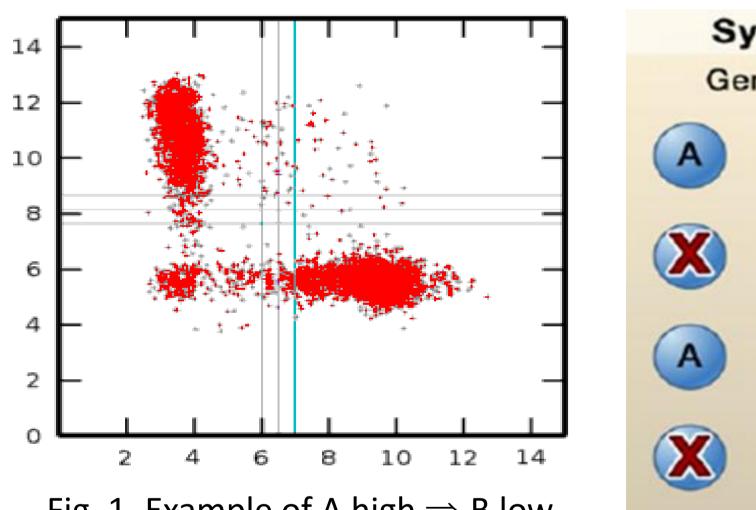


Fig. 1. Example of A high ⇒ B low.X, Y axes are expression level of gene A and B.

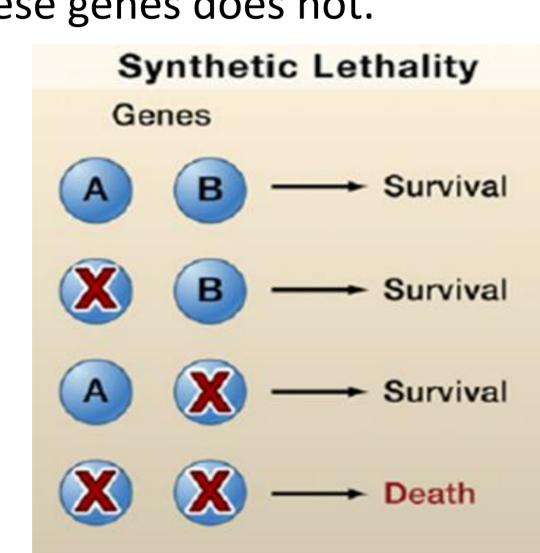


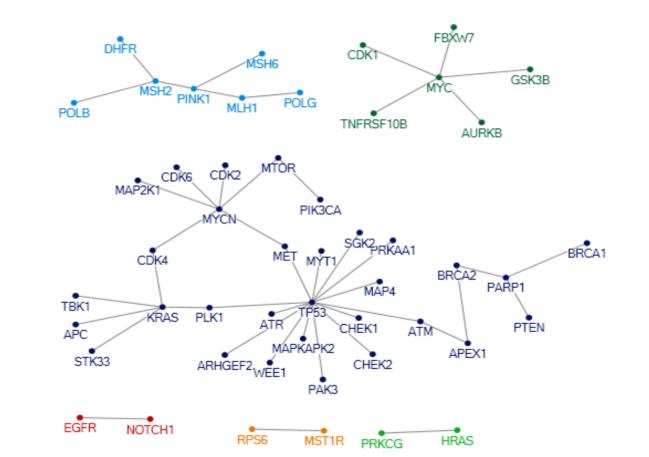
Fig. 2. Synthetic lethality [2]

- The inhibition of synthetic lethal partners of a frequently altered gene in cancer can selectively kill the cancer cell.
- HILO implication (mutual exclusion) has to exist between the alteration of two genes that have a SL interaction.
- Large cancer sequencing projects such as TCGA can be mined to find genetic alterations that are mutually exclusive with a particular genetic alteration.

Methodology and Results

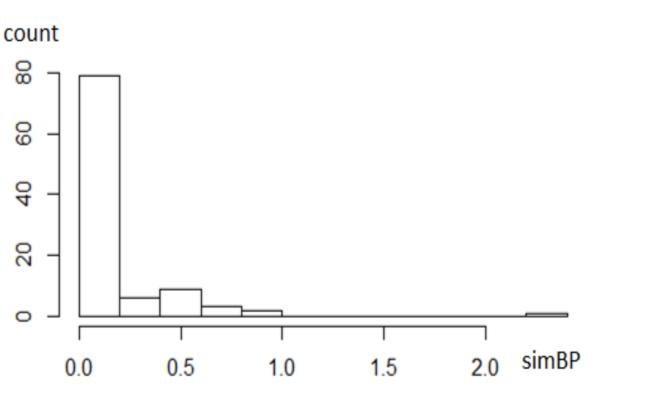
Understanding Known Synthetic Lethal Relations

- Identify informative features that can distinguish SL gene pairs from the others.
- We've collected and plotted 45 known synthetic lethal relationships from literature(Fig. 3).



1. Analysis using Boolean Implication Fig. 3. 45 known SL relations.

- HILO relationship means that two genes are rarely mutated together, thus exhibit mutual exclusion relation, which is expected between SL gene pairs.
- We examined the 45 relations, and found that 10 of them are HILO relationships, which shows HILO implications are useful for predicting SL relations.
- Many of the genes in the 45 pairs had low deletion frequency in the BRCA dataset and hence could not be detected using a statistical test.



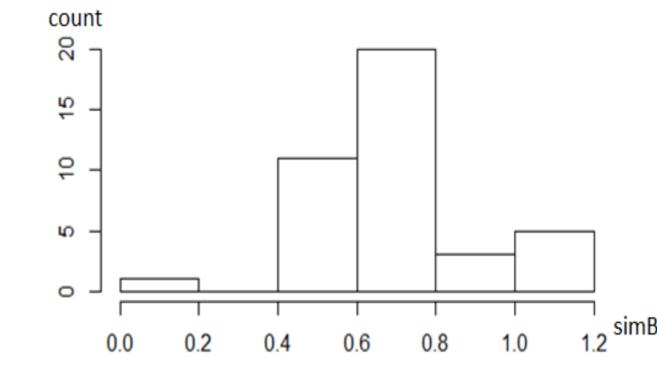


Fig. 4. Dist. of 100 similarity scores between Fig. 5. Dist. of known SLs similarity scores. random gene pairs with HILO relation.

2. Analysis using Gene Similarity

- We analyze these gene pairs by examining their similarity using Lin's method[3].
- Fig. 4 and Fig. 5 are distributions of similarity scores in non-SLs and SLs. We can see a great difference in the peak between the two histograms. Thus a threshold of 0.5 can be used to roughly distinguish between them.

3. Analysis using Pathways Information

- We have searched KEGG pathways that the SL gene pair belong to at the same time.
- Out of 45 known SLs, 4 pairs have more than 3 common pathways, 11 have 1~2 in common, while the rest 30 don't have those in common.

Methodology and Results

Predicting New SL Relations

 According to the analysis, we use Boolean Network and similarity to predict SLs.

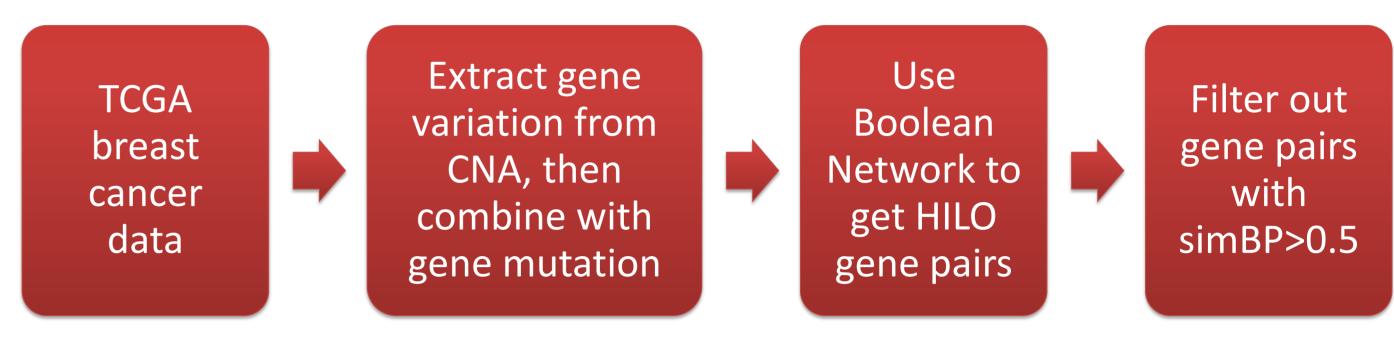


Fig. 6. Flow chart of SL prediction pipeline

Prediction of important genes in cancer

Gene	Alternations num	HILO num	Predicted SL num
TP53_M_mut	155	2051	310
TP53_N_mut	34	1315	197
TP53_FSD_mut	33	914	139
TP53_SS_mut	18	291	39
BRCA1-del	61	291	173
MYC-amp	290	10	7

Fig. 7. Prediction summary of gene TP53, BRCA1 and MYC

Conclusions and Future Work

Conclusions

- HILO implications derived from large cancer sequencing projects such as TCGA can be used to identify novel synthetic lethal relations.
- Functional buffering between SL relation can be captured by Lin's similarity measure[3].
- The prediction pipeline is generic and can be applied on the other cancer types.

Future Work

- Find more features to capture SL properties. Candidates are: 2hop physical-SL feature, gene expression data, protein-protein interaction network, etc.
- Collect experimentally verified non-SL relations, and analyze the difference compared with SL relationships.
- Develop machine learning algorithm to build a classifier.

References

[1]: D. Sahoo, D. Dill, A. Gentles, R. Tishirani and S. Plevrithis, Boolean implication networks derived from large scale, whole genome microarray datasets, Genome Biology, Oct 2008.

[2]: Alan Ashworth, Christopher J. Lord, and Jorge S. Reis-Filho, *Genetic Interactions in Cancer Progression and Treatment*, Cell 145, April 1, 2011 [3]: Lin D. An information-theoretic definition of similarity, Proceedings of the Fifteenth International Conference on Machine Learning. San Francisco, CA, USA: Morgan Kaufmann Publishers Inc.; 1998. p. 296-304

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Further Info

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