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A Bioinformatics Approach to
Synthetic Lethal Interactions in
Breast Cancer with Gene
Expression Data

S. Thomas Kelly

a thesis submitted for the degree of
Doctor of Philosophy
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New Zealand.

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Abstract

Background

Synthetic lethal genetic interactions are re-emerging in the post-genomics era due to their potential for use in precision medicine against cancers. Synthetic lethal drug design exploits the functional redundancy of genes disrupted in cancers (including tumour suppressors) to develop specific treatments against them. *CDH1*, which encodes [E-cadherin](#), is a tumour suppressor gene with loss of function in breast and stomach cancers. Experimental screens have identified candidate synthetic lethal interactions for drug target triage, which can be further supported with bioinformatics analysis. Furthermore, gene expression data enables investigation of synthetic lethal pathways and graph structure of synthetic lethal genes within them.

Approach

A computational methodology, the Synthetic Lethal Prediction Tool ([SLIPT](#)) has been developed to detect synthetic lethal interactions in gene expression data. This methodology was demonstrated on interactions with *CDH1* in breast and stomach cancer data from The Cancer Genome Atlas ([TCGA](#)) project. Synthetic lethal genes and pathways were further investigated with unsupervised clustering, gene set over-representation analysis, metagenes, and permutation resampling. In particular, analyses focused on comparing [SLIPT](#) gene candidates to an experimental [siRNA](#) screen [Telford *et al.* \(2015\)](#). Network analysis methods were applied to the most supported pathways to test for pathway structure among between synthetic lethal candidates. Simulation and modelling was used to assess the statistical

performance of [SLIPT](#), including simulated data with correlation structures derived from graph structures.

Findings

Many candidate synthetic lethal partners of *CDH1* were detected in both [TCGA](#) breast cancer. These genes clustered into several distinct groups, with distinct biological functions and elevated expression in different clinical subtypes. While the number of genes detected by both approaches was not significant, these contained significantly enriched pathways. In particular, $G_{\alpha i}$ signalling, cytoplasmic microfibres, and extracellular fibrin clotting were robustly supported by both approaches, which is consistent with the known cytoskeletal and cell signalling roles of [E-cadherin](#) and validation of [GPCR](#) pathways performed by [Telford et al. \(2015\)](#). Many of these pathways were replicated in stomach cancer data. The pathways supported only by [SLIPT](#) included regulation of immune signalling and translational elongation which were not expected to be detected in an isogenic cell line model but are still candidates for further investigation.

Synthetic lethal candidates detected by [SLIPT](#) and [siRNA](#) were compared within graph structures of the candidate synthetic lethal pathways. These genes did not differ with respect to network metrics of importance or connectivity in the pathway. There was little support, across pathways, that [SLIPT](#) gene candidates were consistently upstream or downstream of [siRNA](#) gene candidates within pathways.

A model of synthetic lethality was used to simulate gene expression data with synthetic lethal partners of a gene. The [SLIPT](#) methodology had high statistical performance, detecting few synthetic lethal partners, which diminished with more synthetic lethal partners or lower sample size. The [SLIPT](#) methodology performed better than Pearson correlation or the χ^2 -test. In particular, it performed well with high specificity for datasets containing thousands of genes or genes positively correlated with the query gene (as expected to occur in expression data). [SLIPT](#) was robust across correlation structures, including those derived from complex pathway structures and often distinguished synthetic lethal genes from those positively or negatively correlated with them. Therefore [SLIPT](#) is appropriate to identify

synthetic lethal genes within pathways and use candidate synthetic lethal genes (and their correlates) to identify synthetic lethal pathways.

Summary

Thus this thesis has developed, evaluated, and refined a bioinformatics approach to discovery of synthetic lethal genes solely from gene expression data. This approach has been demonstrated to detect biologically informative and clinically relevant candidate partners for *CDH1* in breast and stomach cancers. These investigations have also involved the development of network analysis and simulation procedures which may be more widely applicable.

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どうもありがとう由ちゃん。頑張った!もうすぐ行きます。!また来月!

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Glossary

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| allele | A gene variant with a specific sequence and phenotype. |
| bioinformatics | Statistical or computational approaches to biological data or research tools. |
| bisulfite-Seq | Epigenomic data from sequencing bisulfite treated DNA. |
| CAGE-Seq | Transcriptome data from cap analysis of gene expression. |
| cancer | A class of diseases, formally “malignant neoplasm”, of abnormal cellular growth and spread to other organs. |
| cancer gene | A gene which is involved in the malignancy of some cancers, encompassing oncogenes and tumour suppressors , which have molecular aberrations in cancer or variants which predispose individuals to cancer. |
| centrality | A network metric which identifies important vertices . |
| chemoprevention | The use of drugs to prevent early-stage cancers, generally applied to high-risk mutation carriers. |
| chemotherapy | The use of cytotoxic drugs to treat cancers, in combinations, generally applied to advanced stage cancers. |
| ChIP-Seq | Epigenome data from chromatin immunoprecipitation sequencing. |
| compound screen | A high-throughput screen performed using a library of chemical compounds. |
| computational biology | Applying computational or mathematical modelling to understanding biological systems and relationships. |

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| conditional essentiality | A gene becoming essential to viability under certain environmental conditions, including presence of compounds which inactivate other genes. |
| copy number | The number of copies of DNA, typically two copies for diploid organisms but subject to variation. |
| <i>de novo</i> | A bioinformatics sequence assembly conducted entirely from raw genomics data without a reference sequence. |
| diagnosis | The identification of disease by clinical, cellular, and molecular characteristics. |
| driver mutation | A mutation which promotes cancer growth. |
| E-cadherin | Epithelial cadherin (calcium-dependent adhesion), a cell-adhesion protein encoded by <i>CDH1</i> . |
| edge or link | A relationship connecting a pair of elements of a graph structure or network, may be weighted or directional. |
| epigenome | An analysis of epigenetic modifications of all genes in the genome. |
| epistasis (biological) | The effects of a gene modifying or masking the phenotype of another gene. |
| epistasis (statistical) | A divergence of the observed double mutant phenotype from that expected based on the respective phenotypes of single mutant (Fisher, 1919). |
| essential | A gene which is required to be functional or expressed for a cell or organism to be viable, grow or develop. |
| exome | A sequencing approach designed to generate data enriched for coding genes within the genome. |
| familial | A trait recurrently occurring in families, not necessarily with a genetic cause. |
| functional redundancy | Genes which perform a common function, also known as genetic redundancy. |

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| gene expression | A measure of the relative expression of each gene from the mRNA extracted from (pooled) cells. |
| genetic robustness | A system of biological pathways which (has evolved to) continue to function as a whole under various conditions, including the inactivation of various individual genes. |
| genome | All of the DNA sequence in the genome. |
| genomic | The use of data from all genes in the genome. |
| genomic medicine | The use of genomic information to tailor medicine treatment to the genetics of an individual. |
| germline mutation | A mutation that occurred in germline cells and is passed between generation. |
| graph or network | A mathematical structure modelling or depicting the relationships between elements. |
| hallmark of cancer | An underlying characteristic of cancer as part of a rational approach devised by (Hanahan and Weinberg, 2000). |
| hereditary | A trait or disease which has a genetic cause and is inherited from family members. |
| high-throughput screen | An experimental procedure to perform a large scale series of chemical, genetic, or pharmacological tests. |
| hub | A central or highly connected component of a network. |
| <i>in silico</i> | An investigation conducted using computations, typically simulations or analyses. |
| <i>in vitro</i> | An investigation conducted using a controlled experimental system to examine biomolecules. |
| <i>in vivo</i> | An investigation conducted using in the context of a biological cell or organism, including pre-clinical models and clinical trials. |
| induced essentiality | A gene becoming essential to viability under certain conditions, including inactivation of a synthetic lethal partner. |

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| information centrality | A network centrality metric which uses the impact of removing a vertex or node on connections in the network. |
| intrinsic subtype | Distinguishing cancer by molecular and genetic features. |
| MCF10A cell line | A non-tumorigenic epithelial cell line derived from breast tissue. |
| metabolome | All the metabolites and enzymes in the cell. |
| metagene | A consistent signal of expression for a collection of genes such as a biological pathway, derived from singular value decomposition. |
| metagenome | All of the genes and genomes in a community. |
| metastasis | A secondary growth of a tumour or spread of cancer to other organs. |
| methylation | A measure of the epigenetic regulation of DNA at CpG dinucleotide (CpG) sites. |
| microarray | A high-throughput technique to measure presence or abundance of nucleic acid sequences from binding to probes. |
| microRNA | Short RNA molecules generally regarded to regulate gene expression by binding to mRNA. |
| molecular profile | A combination of genetic and biochemical measures which identifies characteristic traits of a tumour. |
| molecular subtype | A classification of cancers based on an identification using molecular properties. |
| mutant | A variant or dysfunctional phenotype arising from a mutation in a gene. |
| mutation | A change in DNA sequence that disrupts gene function. |
| network biology | The application mathematical and computational approaches to networks in understanding biological relationships. |
| network medicine | The use of network biology to understand, prevent, or treat diseases. |
| non-oncogene addiction | The dependence of a cancer cell on functioning non-mutant genes. |
| 'omics | A combination of approaches to generating biological data with high-throughput procedures such as genomics, proteomics or metabolomics. |

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| oncogene | A gene that potentially causes cancer, typically by over-expression or mutant gene variants. |
| oncogene addiction | The dependence of a cancer cell on a specific oncogenic pathway. |
| PageRank centrality | A network centrality metric which uses eigenvectors with a scaling factor (Brin and Page, 1998). |
| pan cancer | A focus on the molecular and genetic features across cancers in different tissues. |
| passenger mutation | A mutation that occurs in cancers but does not affect the growth of cancers. |
| pathway | A series of biomolecules that produces a particular product or biological function. |
| pleiotropy | When a gene has multiple biological functions. |
| polypharmacology | The design of drugs to target multiple molecular targets or biological pathways. |
| precision medicine | The application of prevention and treatment measures to target diseases by molecular and genetic features. |
| prognosis | The estimation of disease progression and patient outcome. |
| proteome | All the proteins expressed from the genome. |
| proto-oncogene | The non-mutant variant or precursor to a mutant oncogene . |
| recurrent mutation | The repeated occurrence of mutations in a particular gene across cancers. |
| RNAi screen | A high-throughput screen performed using a RNA interference (RNAi) . |
| RNA-Seq | The generation of transcriptome data from sequencing RNA. |
| Sanger sequencing | A dideoxy chain termination method for DNA sequencing (named after Fred Sanger). |
| scale-free | A property of a network which has a power law vertex degree distribution, that is several highly connected hub genes and many with very few connections. |

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| shortest path | A path with the fewest possible edges which connects two particular vertices . |
| small world | A property of a network which is highly connected and has a low characteristic path length, derived from the mean shortest path length across all pairs of nodes. |
| somatic mutation | A mutation that occurs in somatic cells, during a patient's lifespan. |
| sporadic cancer | Cancers which do occur in patients with a family history or carry a high-risk genetic variant. |
| synergy | When multiple drugs have more effect than expected from the effect of each separately. |
| synthetic dosage lethal | A synthetic genetic interaction (SGI) analogous to synthetic lethality where where one gene is inactivated and the other over-expressed. |
| synthetic lethal | Genetic interactions where inactivation of multiple genes is inviable (or deleterious) which are viable if inactivated separately. |
| synthetic lethal screen | A high-throughput screen performed on isogenic cell lines to detect genes for which inhibition specifically deleterious to the null mutant genotype. |
| synthetic rescue | A synthetic genetic interaction when the combined mutations restores the wild-type the phenotype of one of the mutations . |
| synthetic sick | Genetic interactions where inactivation of multiple genes is deleterious which are viable if inactivated separately. |
| synthetic suppression | A synthetic genetic interaction when the combined mutations (partially) suppresses the mutant phenotype of one of the mutations . |
| targeted therapy | Cancer treatment that specifically acts against a molecular target, in contrast to standard chemotherapy. |
| transcriptome | All of the genes expressed in the genome. |
| treatment | Medical procedures for a disease to improve patient outcomes. |
| tumour | An abnormal lump of tissue or growth of cells, may be cancerous. |

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| tumour suppressor | A gene potentially causes cancer, typically by disruption of functions which protect the cell from cancer. |
| vertex degree | A network metric of connectivity of vertices which uses the number of edges connected to each vertex or node . |
| vertex or node | An element of a graph structure or network. |
| wild-type | A natural phenotype of a trait or the normally functional allele which encodes it. |

Acronyms

| | |
|----------|---|
| 1KGP | 1000 genomes project. |
| ADP | Adenosine Diphosphate. |
| AMP | Adenosine Monophosphate. |
| AMPK | AMP -activated Protein Kinase. |
| ANOVA | Analysis of Variance. |
| ATP | Adenosine Triphosphate. |
| AUROC | Area Under the Receiver Operating Characteristic (curve). |
| BioPAX | Biological Pathway Exchange. |
| BiSep | Bimodal Subsetting Expression. |
| BMP | Bone Morphogenic Protein. |
| cAMP | Cyclic AMP . |
| CCL | Cancer Cell Line Encyclopaedia. |
| cDNA | Complementary DNA (from mRNA). |
| CGP | Cancer Genome Project. |
| ChIP | Chromatin Immunoprecipitation. |
| ChIP-Seq | Chromatin Immunoprecipitation Sequencing. |
| CNV | Copy Number Variation. |
| COSMIC | Catalogue Of Somatic Mutations In Cancer. |
| CpG | 5'-C-phosphate-G-3'. |
| CRAN | comprehensive R archive network. |
| CXCR | Chemokine Receptor. |
| DAISY | Data Mining Synthetic Lethal Identification Pipeline. |
| DDBJ | DNA Data Bank of Japan. |
| DNA | Deoxyribonucleic Acid. |
| EMBL | European Molecular Biology Laboratory. |
| EMT | Epithelial-Mesenchymal Transition. |
| ENA | The European Nucleotide Archive. |
| ENCODE | Encyclopaedia of DNA Elements. |

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| ER | Estrogen Receptor. |
| exprSL | Synthetic Lethality (expression). |
| FANTOM | Functional Annotation Of Mammalian genome. |
| FDR | False Discovery Rate. |
| GEO | Gene Expression Omnibus. |
| GO | Gene Ontology. |
| GPCR | G Crotein Coupled Receptor. |
| HDAC | Histone Deacetylase. |
| HDGC | Hereditary Diffuse Gastric Cancer. |
| HLRCC | Hereditary Leiomyomatosis and Renal Cell Carcinoma. |
| HPC | High Performance Computing. |
| ICGC | International Cancer Genome Consortium. |
| InDel | Insertion or Deletion (in DNA sequence). |
| JAK | Janus Kinase. |
| lncRNA | Long Non-Coding RNA. |
| METABRIC | Molecular Taxonomy of Breast Cancer International Consortium. |
| microRNA | Micro RNA. |
| mRNA | Messenger RNA. |
| MSI | Microsatellite Instability. |
| mtSL | synthetic Lethality (mutation). |
| mtSLIPT | Synthetic Lethal Interaction Prediction Tool (against mutation). |
| NCBI | National Center for Biotechnology Information (in the USA). |
| NCI | National Cancer Institute (in the USA). |
| NeSI | New Zealand eScience Infrastructure. |
| NGS | Next-Generation Sequencing. |
| NHGRI | National Human Genome Research Institute (in the USA). |

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| NIG | National Institute of Genetics (in Japan). |
| NIH | National Institutes of Health (in the USA). |
| NMD | Nonsense-Mediated Decay. |
| PAM50 | Prediction Analysis of Microarray 50. |
| PARP | Poly-ADP-Ribose Polymerase. |
| PCR | Polymerase Chain Reaction. |
| PDE | Phosphodiesterase. |
| PI3K | Phosphoinositide 3-kinase. |
| PIP ₂ | Phosphatidylinositol-(4,5)-bisphosphate. |
| PIP ₃ | Phosphatidylinositol-(3,4,5)-trisphosphate. |
| PPI | Protein-Protein Interaction. |
| PR | Progesterone Receptor. |
| qPCR | Quantitative (real-time) Polymerase Chain Reaction. |
| RFLP | Restriction Fragment Length Polymorphism. |
| RGS | G-protein Signalling. |
| RHO | Ras Homolog Family. |
| RMA | Robust Multiarray Averaging (normalisation). |
| RNA | Ribonucleic Acid. |
| RNAi | RNA Interference. |
| ROC | Receiver Operating Characteristic (curve). |
| RPPA | Reverse Phase Protein Arrays. |
| RRBS | Reduced Representation Bisulfite Sequencing. |
| rRNA | Ribonucleic acid. |
| RSEM | RNA-Seq by Expectation Maximization (normalisation). |
| SGA | Synthetic Gene Array (technique). |
| SGI | Synthetic Genetic Interaction. |
| shRNA | Short Hairpin RNA. |
| siRNA | Short Interfering RNA. |
| SL | Synthetic Lethal. |
| SLIPT | Synthetic Lethal Interaction Prediction Tool. |
| Slurm | Simple Linux Utility for Resource Management. |
| SNP | Single Nucleotide Polymorphism. |
| SR | Synthetic Rescue (or viability). |
| SS | Synthetic Suppression. |

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| SSL | Synthetic Sick. |
| TCGA | The Cancer Genome Atlas (genomics project). |
| TGF α | Transforming Growth Factor α . |
| tRNA | Transfer RNA. |
| UCSC | University of California, Santa Cruz. |
| UTR | Untranslated Region (of mRNA). |
| WNT | Wingless-Related Integration Site. |

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