Inference of logic networks from insertion and expression data Jeroen de Ridder^{1,2,*}, Marcel Reinders¹ and Lodewyk Wessels^{1,2} * J.deRidder@TUDelft.nl

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Abstract

In this study, 43 tumors, that were induced by retroviral insertional mutagenesis, are profiled, resulting in a dataset for which both the initiating events (the viral integration sites) as well as the consequent expression profiles are available.

We infer associations between insertion loci and gene expression profiles, while explicitly incorporating simple boolean logic, modelling multiple and parallel oncogenic pathways. We show that this results in the discovery of interesting causal associations between virally inserted loci and differentially expressed genes in tumorigenesis.

Problem description





t-max

Virally induced tumors are harvested and expression profiles as well as insertion profiles are determined

Associations between individual insertion loci and gene expression profiles may not be identifiable. This can be explained by the fact that often multiple viral triggers are required for tumorigenesis. Alternatively, parallel pathways lead to multiple possibilities in which viral triggers can disrupt healthy cell proliferation.

Therefore, we infer association between small logic networks consisting of AND and xor functions.

Association is measured by means of the t-score.

Since the data consists of 112 insertion loci in 43 tumors, and expression measurements for all genes, a challenging search task results. Therefore, we propose a backward branch-and-bound search procedure.

Branch-and-bound

It is desirable to allow some mismatch between \tilde{g} (the binary version) of the gene expression) and the output of the logic network. Therefore we introduce Δt , that, for each tumor, measures the effect of a mismatch on the total t-score.

For this search problem a branch-and-bound search strategy is employed.

p_n contains the rows from the search matrix corresponding to tumor n. In step (a) the AND of all possible combinations of rows in p_1 and p_2 is taken. This result is pruned, removing invalid, and duplicate results (indicated by $p_{1,2}^*$). In step (b) the same is done for all possible combinations of rows from $p_{1,2}^*$ and p3. In case no valid solutions result (e.g. $p_{1,2}^*$ is empty), the search continues directly to p_3 (step (c)).

Whole branches in the search tree can be abandoned, by requiring the error to be below a certain tolerance level. (Exemplified by the branches shaded in red).



Conclusions

We evaluated 17 candidate logic networks for the 1000 genes with highest variance across the tumors. This resulted in 55 genes for which the expression profiles associated with a logic network of the insertion loci with p-value < 10^{-6} .

On the right the top 3 networks are given. For all three networks it holds that a hit in any of the input loci is sufficient to upregulate the expression of the target gene.

In conclusion, by integrally analyzing the combination of insertion data and expression data from the same tumor, it is possible to gain insight in the mechanisms in which viral triggers act on their downstream targets, and, as a result, discover interesting genes for further study.

> Left: result for the *Notch1* gene. Clearly, association is only obtained when individual input loci are combined using the logic circuit (lower left panel). The right panel gives the insertion profiles.

Right: result for Dock9



The optimal binary representation of the continuous expression vector is determined by maximizing the t-score.

Logic network under investigation and its truth table. The rows in the truth table with output '1' (black) apply to the tumors for which \tilde{g} equals '1', and vice versa.



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For each row in the truth table it can be determined which loci need to be selected as input to the logic network in order to reach the required output for a certain tumor. For t_1 , row r_2 requires the inputs i_1 and i_2 to be either I_1 , I_2 and I_4 input i_3 to be I_3 or I_5 .















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