### k-PathA: k-shortest Path Algorithm (pronounce as "Qué Pasa")

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## The Situation



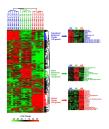
- Situation: Tons of available data on different levels
- Challenge: Derive useful information of structure and behavior of the biological systems

# The Systems-Biology Approach



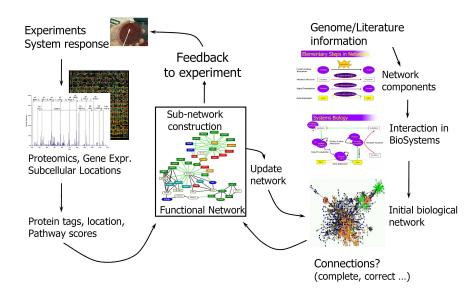
- Before SB: extensive research on single components
- SB says: interactions between components are more important
- SB looks beyond single genes, proteins, etc..
- SB looks at networks because they can give more insights
- SB combines data from different levels, components, conditions

## Gene Expression

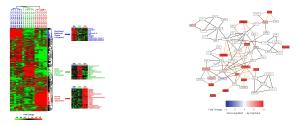


- Cell response to external conditions
- Expression Change = Control vs Experiment
- Biomarkers = Genes with high Expression Change
- Correlated genes are likely to share function
- · Gene expression profiling, Gene clustering show some success
- But ...

### **Response Networks**

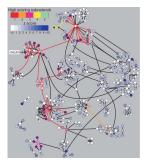


## Gene Clusters vs Response Networks



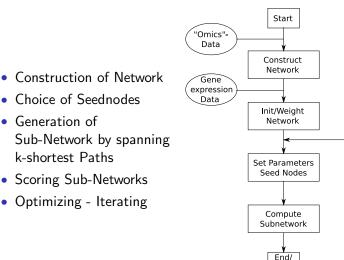
- Genes in Response networks do not have to be active under all conditions
- Response networks underlie the constraints of the molecular interaction network
- Also non-correlated Genes can be included if they connect correlated genes
- Response networks of different conditions can be combined (response networks of new drugs can be compared to those of known drugs)

### Response Networks - Computation



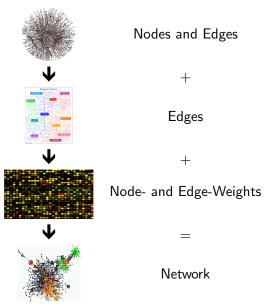
- Generated Sub-Networks have to be scored and compared against random Sub-Networks
- Computationally costly Heuristics
- Still costly for large Networks parallel Heuristics

## k-PathA Approach - Overview

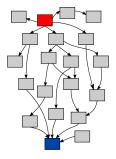


Iterate

### Network Construction

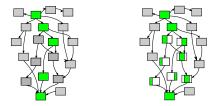


### Subnetworks - Preperation



- Choice of the Sub-Network borders = Seed Nodes
- Option 1: User Input
- Option 2: Randomly
- Option 3: Set of Nodes with highest weights
- In the Optimization step the set is changed

### Sub-Networks - Generation

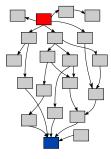


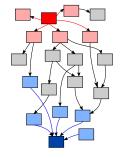
### Definition

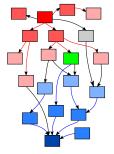
The *k*-th shortest simple path problem consists of the determination of a set of simple (loopless) paths between two nodes  $P^k = \{p_1, p_2, \dots, p_k\} \subseteq P$ , such that  $\forall p \in P - P^k \land p^k \in P^k : cost(p^k) \leq cost(p)$ .

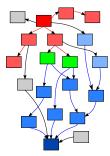
- Spanning k-shortest simple (loopless) Paths
- Forward-, Backward Chaining Approach = Search from both Seeds
- Combination of Paths from both sides

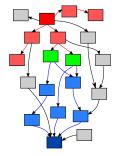
### Sub-Networks - Generation

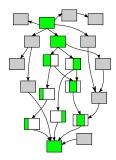












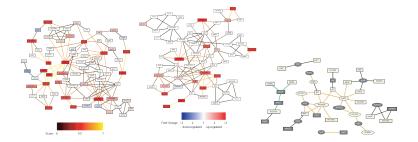
### Sub-Networks - Scoring

- Expression values are normalized
- Nodeweights are  $\geq 0$
- $\bullet \ < 1 \ {\rm down-regulated}$
- $\bullet \ > 1$  up-regulated

• 
$$LinkCost(I_{x,y}) = log(|x+y|)$$

• 
$$PathwayCost(p) = \frac{\sum_{l \in L_p} LinkCost(l)}{|L|^k}$$

### Sub-Networks - Combination

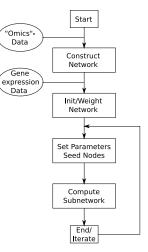


- Combine Sub-Networks for different Conditions
- Can show Similarities of Drug Responses

• 
$$PC(X, Y) = \frac{(E(XY) - E(X)E(Y))}{\sqrt{Var(X)Var(Y)}}$$

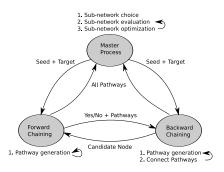
- $FPC(X, Y) = PC(X^{(2,n)}Y^{(1,n-1)})$
- $BPC(X, Y) = PC(Y^{(2,n)}X^{(1,n-1)})$

## Sub-Networks - Optimization



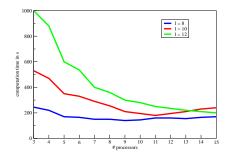
- Change of Seed nodes in every iteration
- Genetic algorithm: e.g. hill climbing

## Parallel Implementation



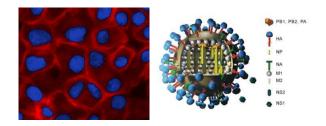
- Communication through message passing
- No shared Memory necessary
- 3 types of Nodes
- Master Sub-Network Handling
- Forward Path search and Common Node Checking
- Backward Path search and Combination of Paths

### Parallel Implementation Behavior



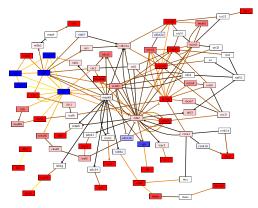
- For larger Networks or high k and I: it scales linear
- For small Networks: communication is big part

## Application



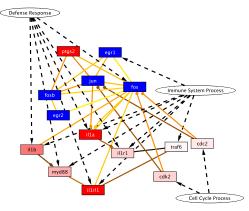
- Human-hybrid Network: 45,041 nodes + 438,567 edges
- Human bronchial cells
- Influenza (H5N1) infection (24h)
- Control: Mock infection (24h)
- k = 3, l = 13

## Application



- Two well known processes affected by viral infections are present in the response network
- Cell-Cycle Genes (cdk2 and cdc's) are up-regulated
- Transcription factors known as early responders (fos and jun) are down-regulated

## Application



- Comparison with other viral infection (LMCV, Djavani et al)
- Cell Cycle Process is modulated to induce apoptosis in the same way
- Immune Response is targeted similarly
- Moderate differences in the Host Response

## Conclusions

- Computationally feasible through Parallel Implementation
- No shared Memory needed
- Flexible because new kinds of interactions can be introduced
- Flexible because of different scoring functions
- Response Networks for different conditions
- Comparison with other drug or viral responses possible

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For your Attention