#### Cancer Gene/Module Identification

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### **Informal Definition**

• Given mutations data (TCGA etc.):

Find driver genes/modules in cancer

- Two problems:
  - Candidate Genes
  - Candidate Modules
- Problem with Gene Ranking:

Mutations at different loci could lead to the same disease

## **Driver Module Identification**

- De novo Methods:
  - Rely only on genetic data
  - [Miller et al. 11; Vandin et al. 11, Leiserson et al. 13, Liu et al. 17]
  - Algorithmically, heavy submatrix type problems
  - Disadvantage:
    - Solution space is large
    - Prefiltering based on frequencies may miss rare mutations

#### **Driver Module Identification**

- Knowledge-based Methods:
  - Interaction networks as well as genetic data
  - Coverage oriented methods:
    - Heat diffusion of mutation frequencies
    - Algorithmically, heavy subgraph type problems
    - Hotnet, Hotnet2, HierHotnet ...
  - Coverage+Mutual Exclusion:
    - Simultaneous mutations of genes in shared pathways not frequent
    - Algorithmically, greedy seed-and-extend type heuristics
    - MEMo, BeWith, MEMCover ...

#### **Driver Module Identification**

• Formally, given S<sub>i</sub> and PPI network G,

$$\begin{split} MEX(M) &= \frac{|\bigcup_{\forall g_i \in M} S_i|}{\sum_{\forall g_i \in M} |S_i|} \\ CO(M) &= \frac{|\bigcup_{\forall g_i \in M} S_i|}{|\bigcup_{\forall g_i \in V} S_i|}. \end{split} \qquad \begin{aligned} MS(P) &= \sum_{\forall M_q \in P} RS(M_q) \times MEX(M_q). \\ CS(P) &= \sum_{\forall M_q \in P} \frac{1 - RS(M_q)}{\sum_{\forall M_t \in P} 1 - RS(M_t)} \times CO(M_q), \end{aligned}$$

- Find P, maximizing  $DMSS(P) = MS(P) \times CS(P)$  such that
  - Subgraph induced by each M is connected
  - Min module size and total size below given thresholds
- Computationally intractable

# Algorithm for Module Identification

- Generate node-weighted edge-weighted graph
  - reflecting coverage and mutual exclusion

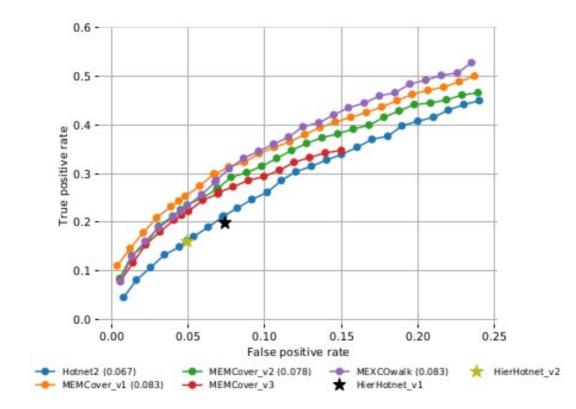
 $w(g_i,g_j) = MEX_n(g_i,g_j) \times CO(\{g_i\}) \times CO(\{g_j\}).$ 

- Edge-weighted random-walk with restart
- Initial modules:
  - strongly connected components
  - similar to other heat-diffusion methods
- Split-and-extend large modules

#### **Evaluation Metrics**

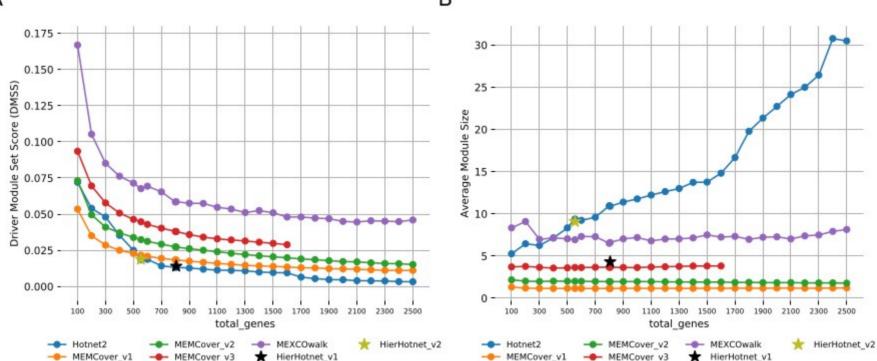
- Systematic evaluations of previous work:
  - Static evaluations based on reference sets (COSMIC etc.)
- Module-specific systematic evaluations missing:
  - Carefully defined optimization scores as in DMSS
  - Cancer type/subtype specificity score
  - Classification (normal vs tumor) accuracy score

## Static Evaluations



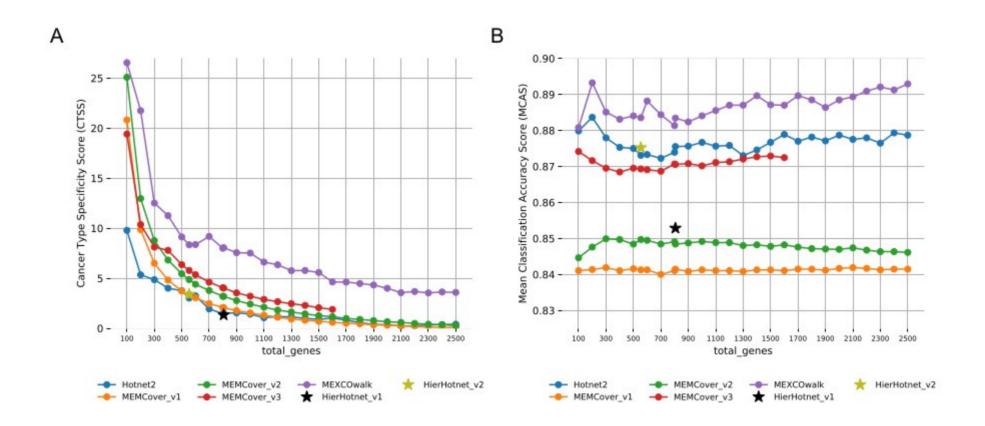
# DMSS and Average Module Size

A

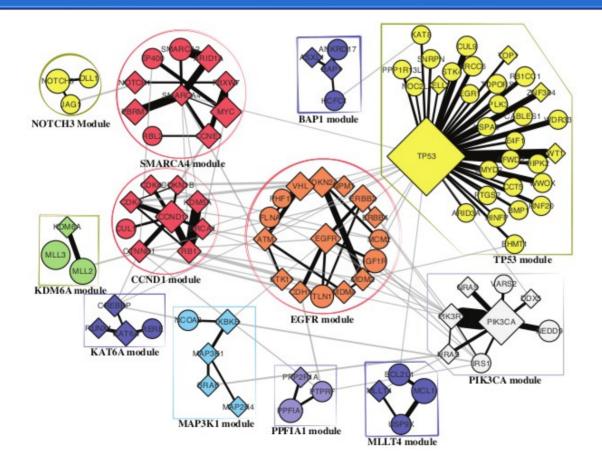


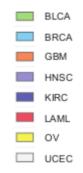
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#### Type Specificity and Classification



#### Modules on Pancancer Data





# Open problems

- Module-oriented
  - optimization problem definitions
  - evaluation criteria
- Especially for the overlapping modules case
- Computational complexity results on sparse graphs



# THANK YOU