Network-based stratification of tumor mutations

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Goal

 Tumor stratification: to divide a heterogeneous population into clinically and biologically meaningful subtypes based on molecular profiles

Previous attempts

Glioblastma and breast cancer – mRNA expression data

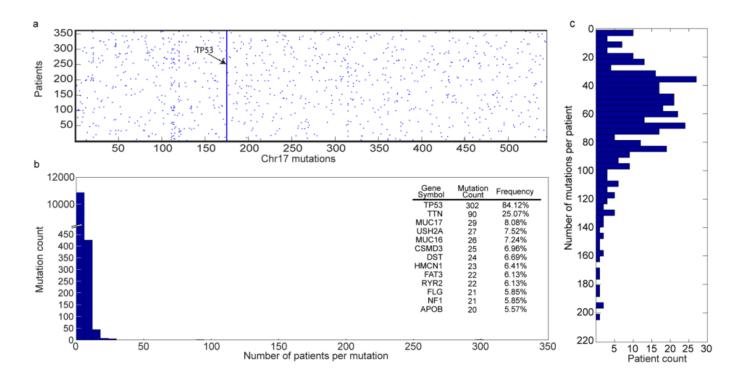
 Colorectal adenocarcinoma and small-cell lung cancer – expression data **not** correlate with clinical phenotype

Somatic mutation profile

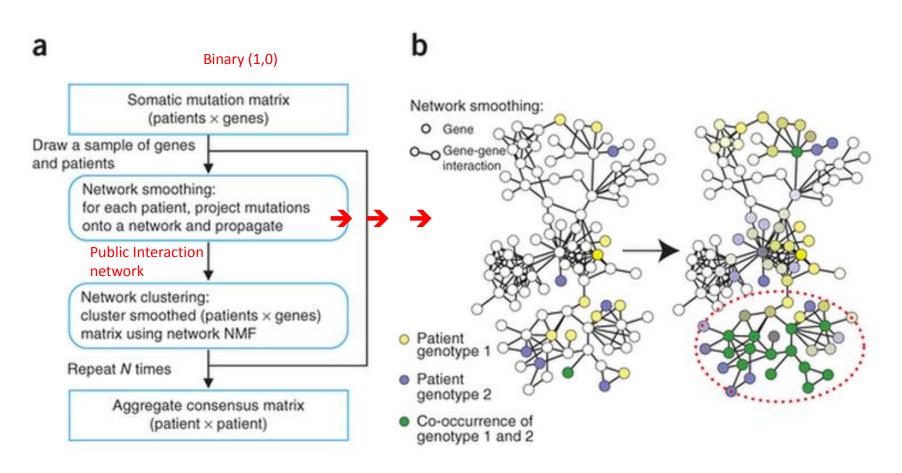
 Compare the genome or exome of a patient's tumor to that of the germ line

Supplementary Figure 1

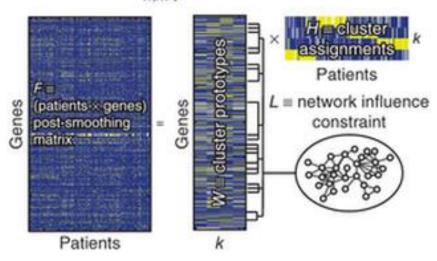
Sparse



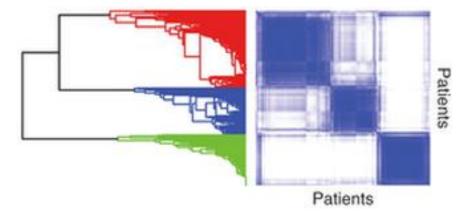
Overview of network-based stratification



C Network NMF: $\min_{W,H>0} |F - WH|I + \gamma |IW^{\dagger}L|I_F$



d Network-based stratification



Network smoothing

• $F_{t+1} = \alpha F_t A + (1-\alpha) F_0$

F₀: patients * genes matrix

A: adjacency matrix of the gene interaction network (STRING, HumanNet and PathwayCommons)

α: tuning factor that determines how far a mutation signal can diffuse

Network-regularized NMF

• Min || F – WH ||² + trace(W^tKW)

Patient * gene matrix

W: a collection of basis vectors, "metagenes"

H: the basis of vector loading

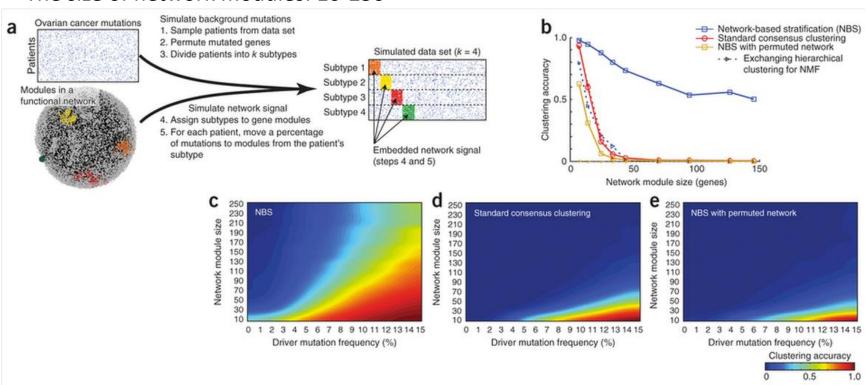
Trace(WtKW): constrain the basis vectors(W) to respect

local network neighborhoods

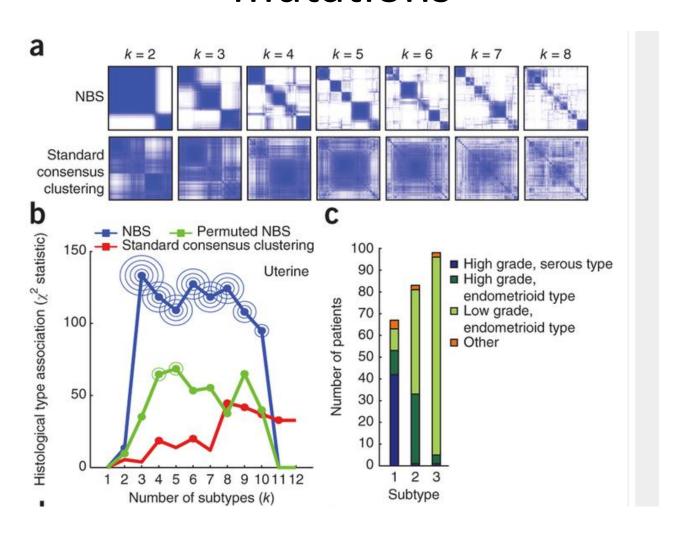
K: derived from the original network

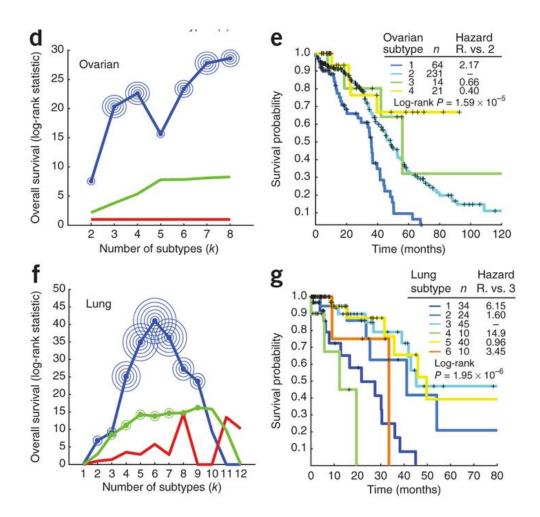
Simulation Assessment

K=4
Driver mutation f: 0% to 15%
The size of network modules: 10-250

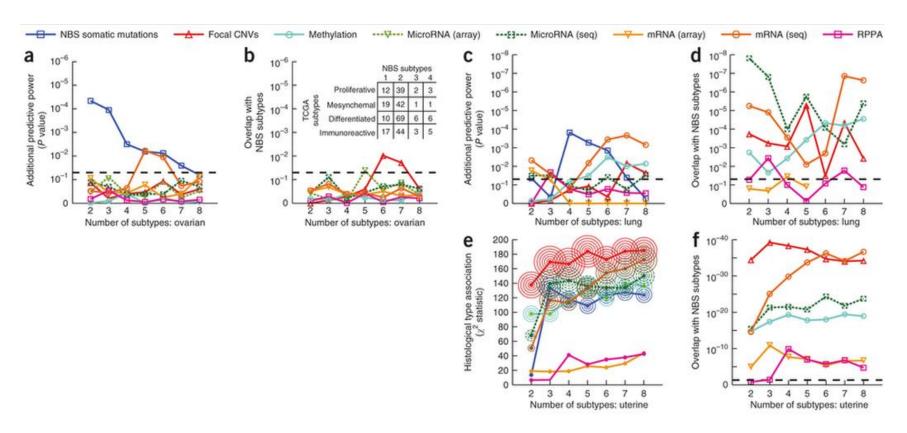


Results- NBS of somatic tumor mutations

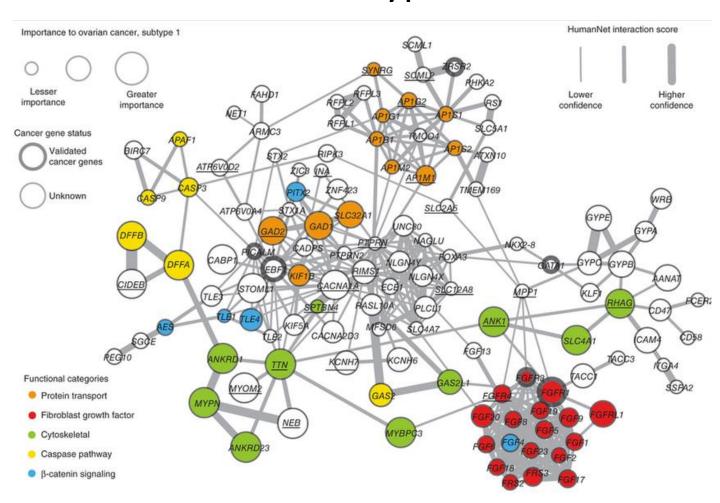




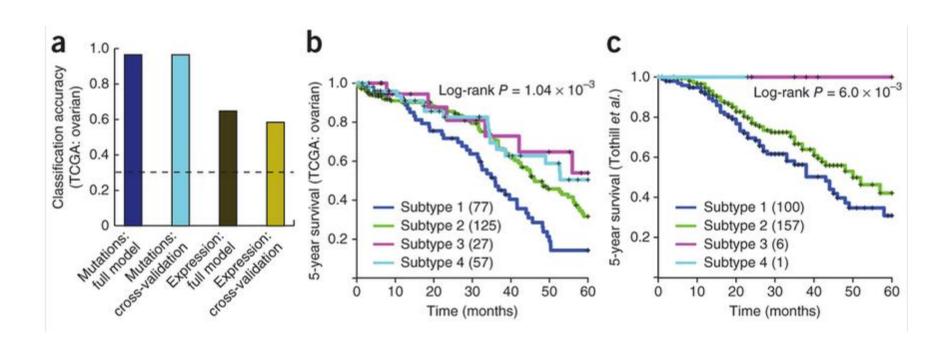
Results-Predictive power and overlap of subtypes derived from different TCGA datasets



Network view of genes with high networksmoothed mutation scores in HumanNet ovarian cancer type 1



From mutation-derived subtypes to expression signatures



Effects of different types of mutations on stratification

