



# Phenotype Switching during Tumor Necrosis Factor alpha Signaling

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## Abstract

Tumor Necrosis Factor alpha (TNF $\alpha$ ) is a pleiotropic cytokine involved in phenotypic decisions such as apoptotic/necrotic death, proliferation. Aberrant TNF $\alpha$  signaling is implicated in numerous pathological conditions. Designing therapeutic strategies to modulate these conditions require insights into the mechanisms governing context-specific phenotypic response to TNF $\alpha$ . Signal transduction culminating in such responses is orchestrated by underlying molecular network of nodes interconnected by edges. Using a comprehensive, well-annotated, manually curated TNF- $\alpha$  signaling network, we show that a graph-theory based dimensionality reduction via modularization can lead to functionally consistent, conserved modules in the network. We identify 19 candidates which when knocked down one-at-a-time significantly disturbs the network robustness yet preserves network modularity. Boolean dynamic simulations and attractor analysis of the underlying state transition graph show that targeting cIAP1/2 and MKRN1 can lead to reliable phenotype switching from proliferation to apoptosis. Knocking off BAX-BAK and LUBAC may result in switching from apoptosis to proliferation. These combinations causing phenotype switching could be potential targets for TNF $\alpha$  based therapeutic strategies.

## Methodology

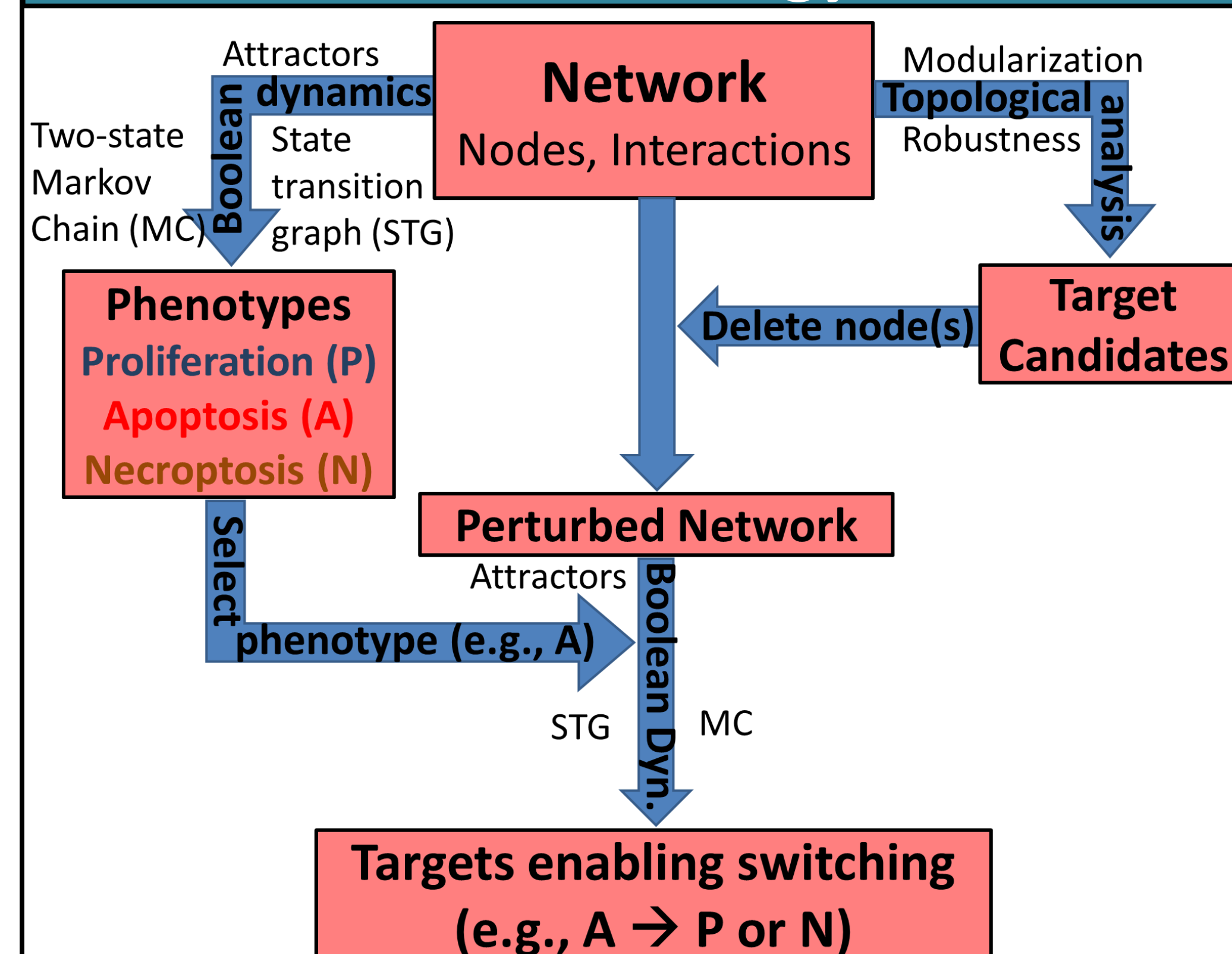


Figure 1: Methodology adopted for identification of targets that enable phenotype switching.

## Network statistics

Attribute	Count
Nodes	423
Edges	341
Proteins	275
Genes+mRNA	62
Phenotypes	4
Primary Literature	663
Boolean rules (Logic)	284
Input + Housekeeping nodes	140
Fixed-point attractors	27

## Manually curated TNF $\alpha$ Signaling Network

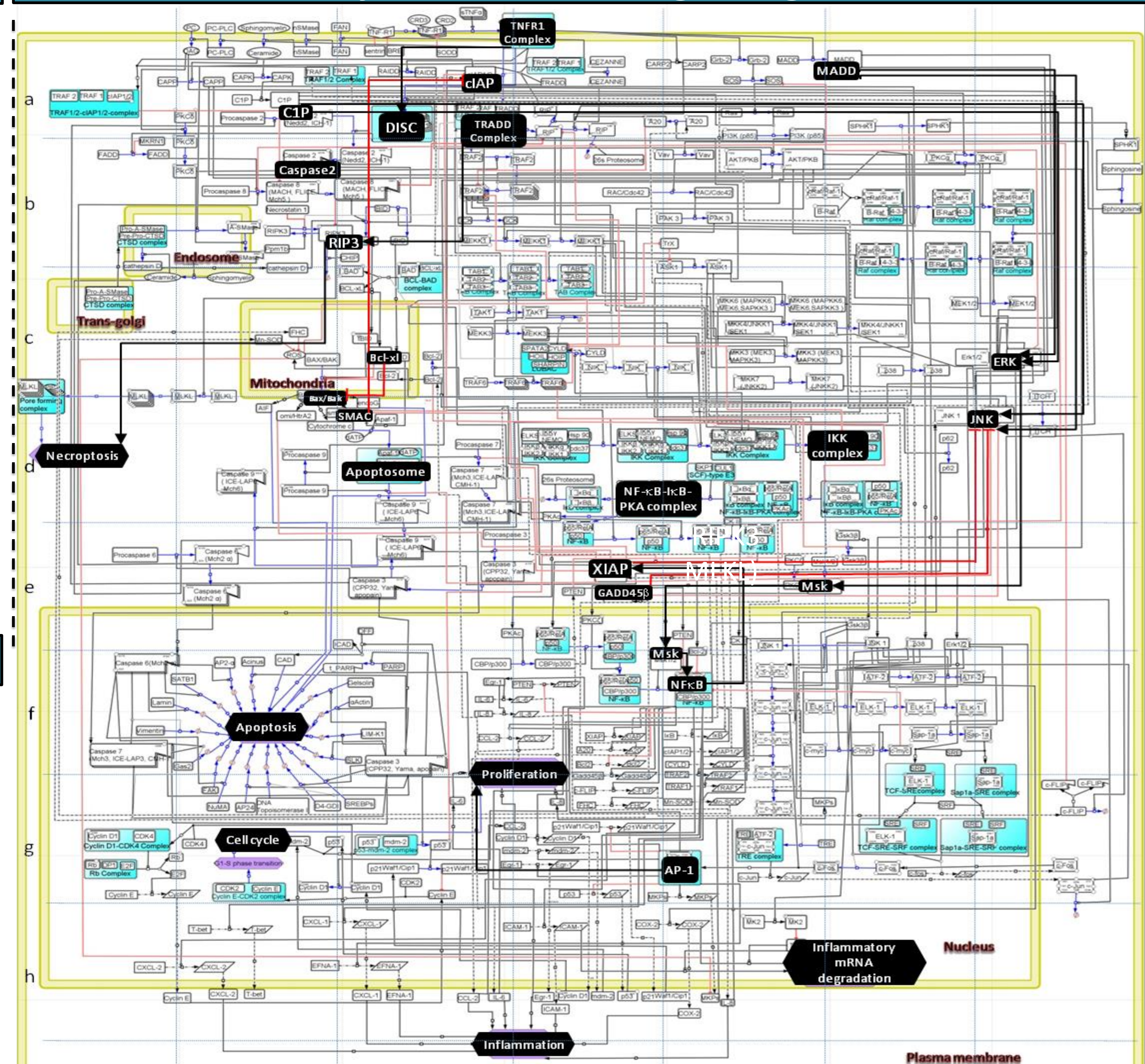


Figure 2: Manually curated comprehensive TNF $\alpha$  signaling network.

## Target identification: Topological and fixed-point attractor analysis

### Network consists of 24 functionally consistent modules

### Phenotype probability

### Phenotype Switching

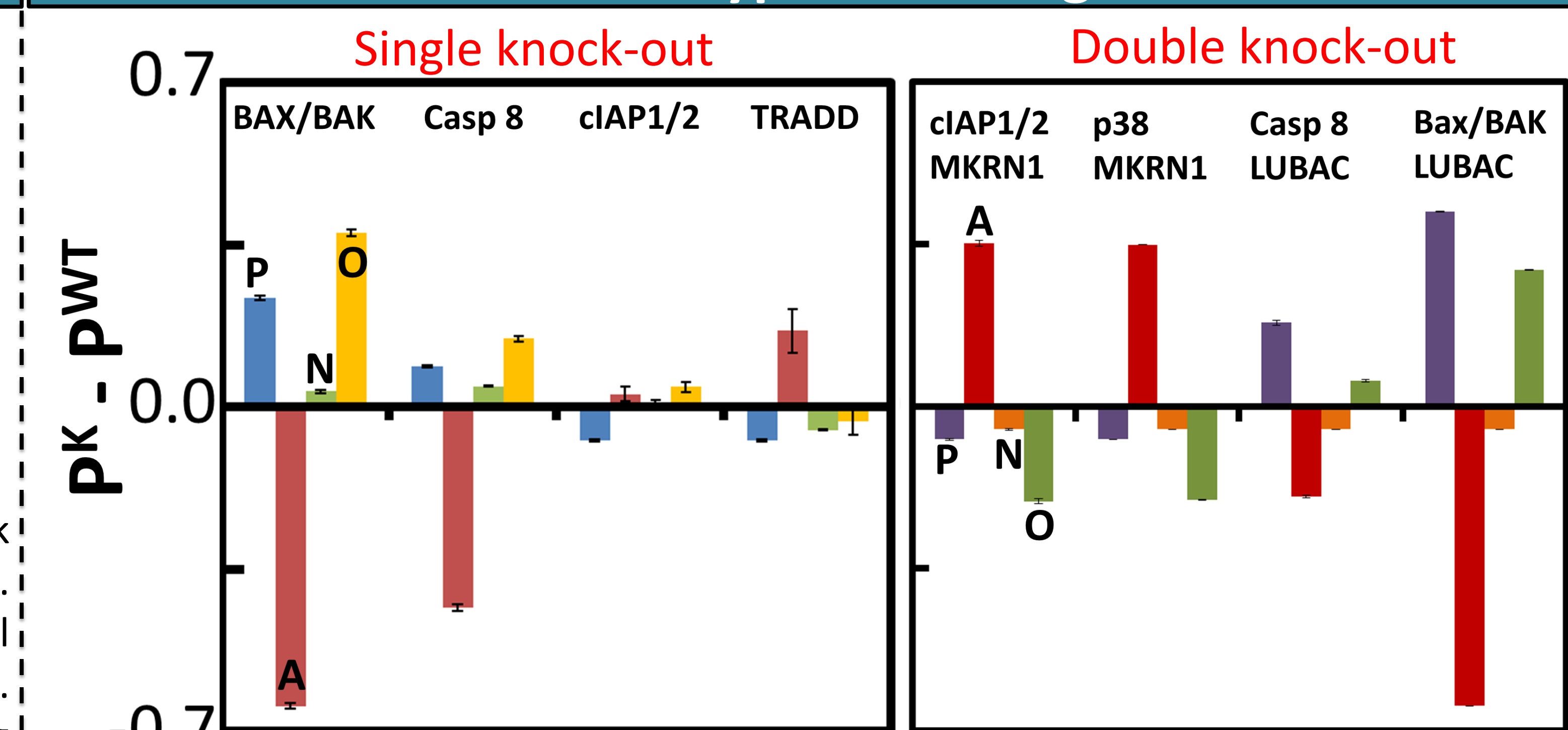
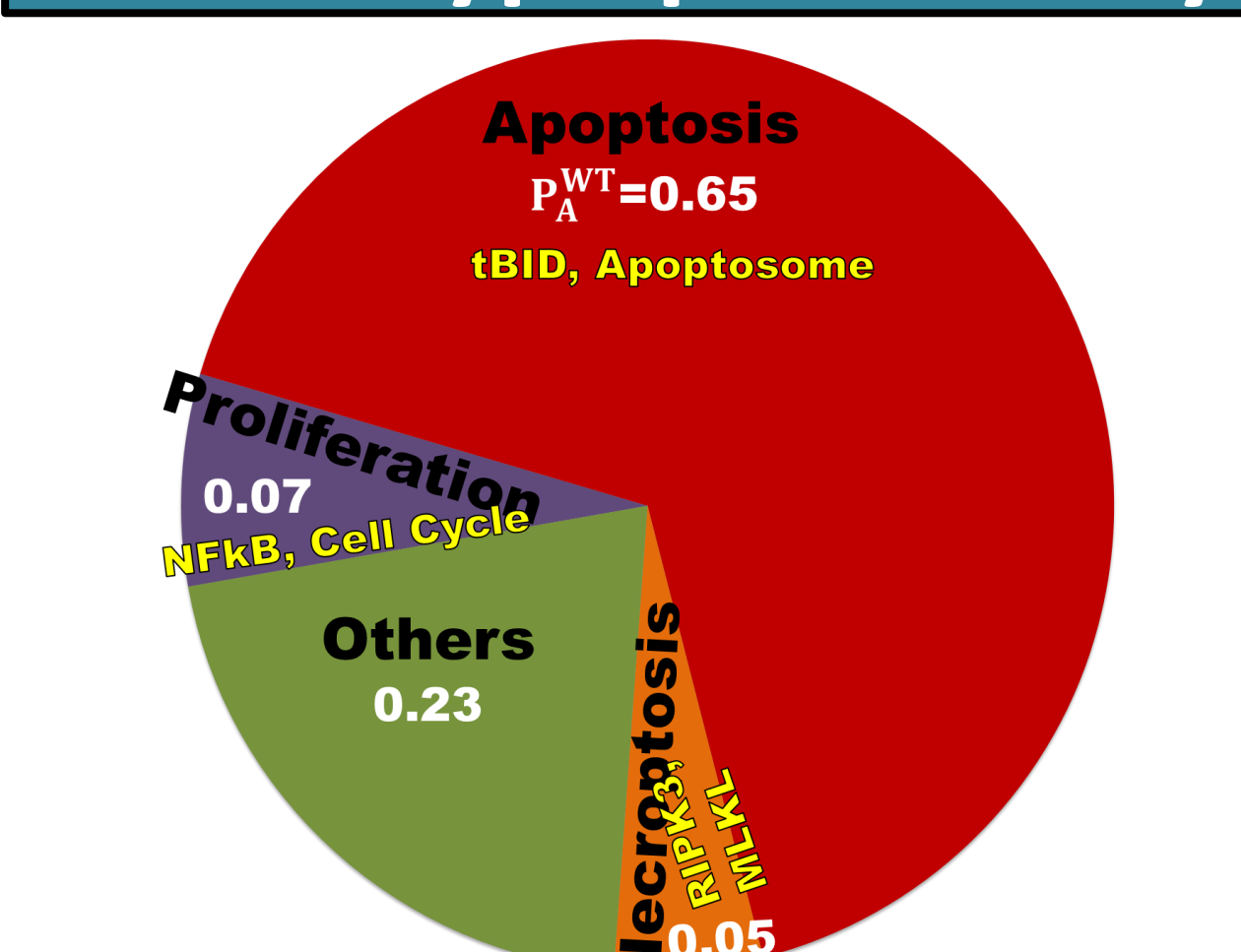
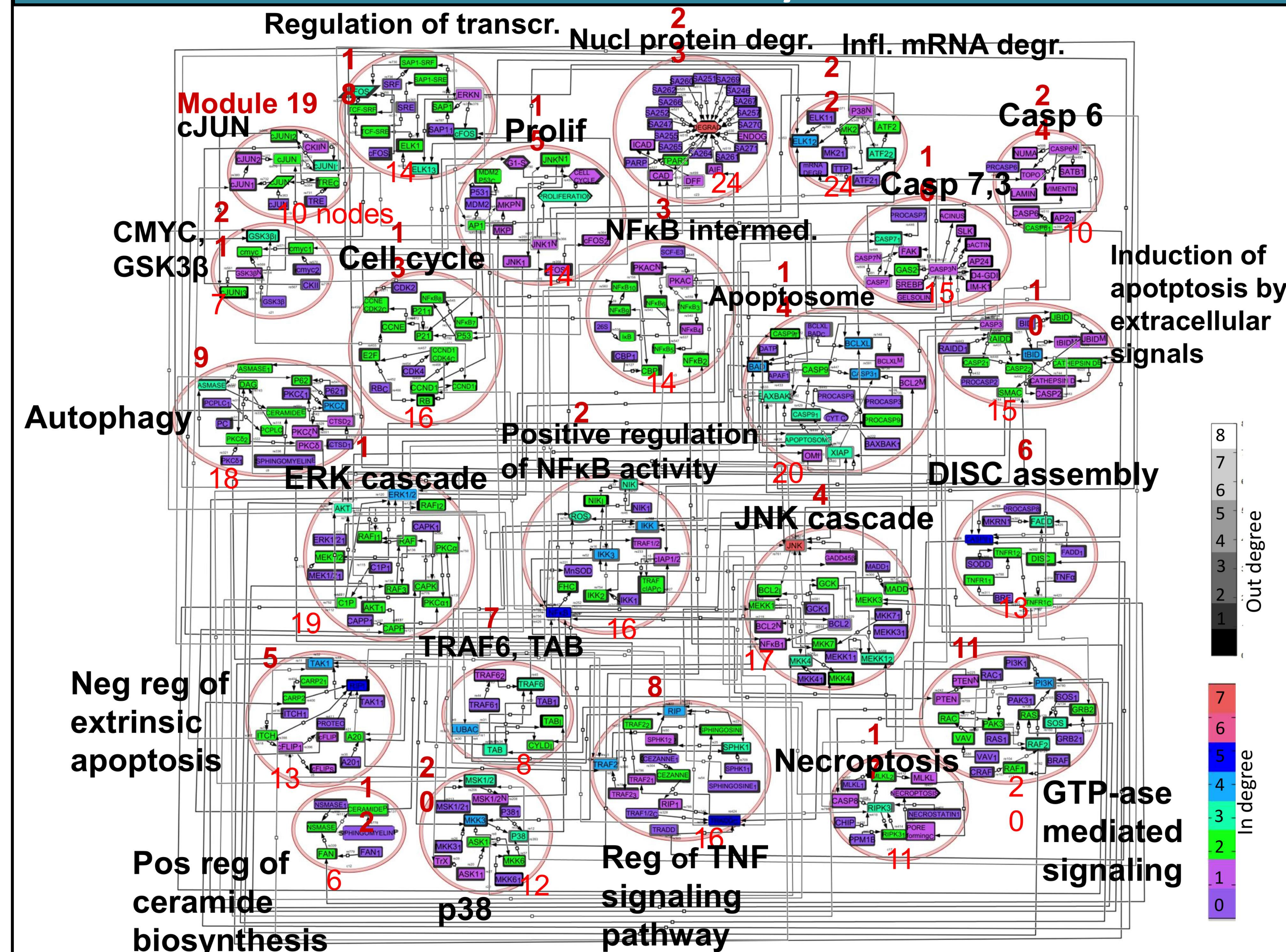


Figure 4: Probability ( $P^{WT}$ ) of network settling into a certain phenotype. WT refers to the original (unperturbed) network in Figure 2. Probabilities were estimated using two-state Markov model.<sup>4,5</sup>

Figure 6: Phenotype switching: Change in the probability of perturbed network settling into a phenotype reflecting the switching caused by the perturbation via single/double knock-out.

### Target candidates

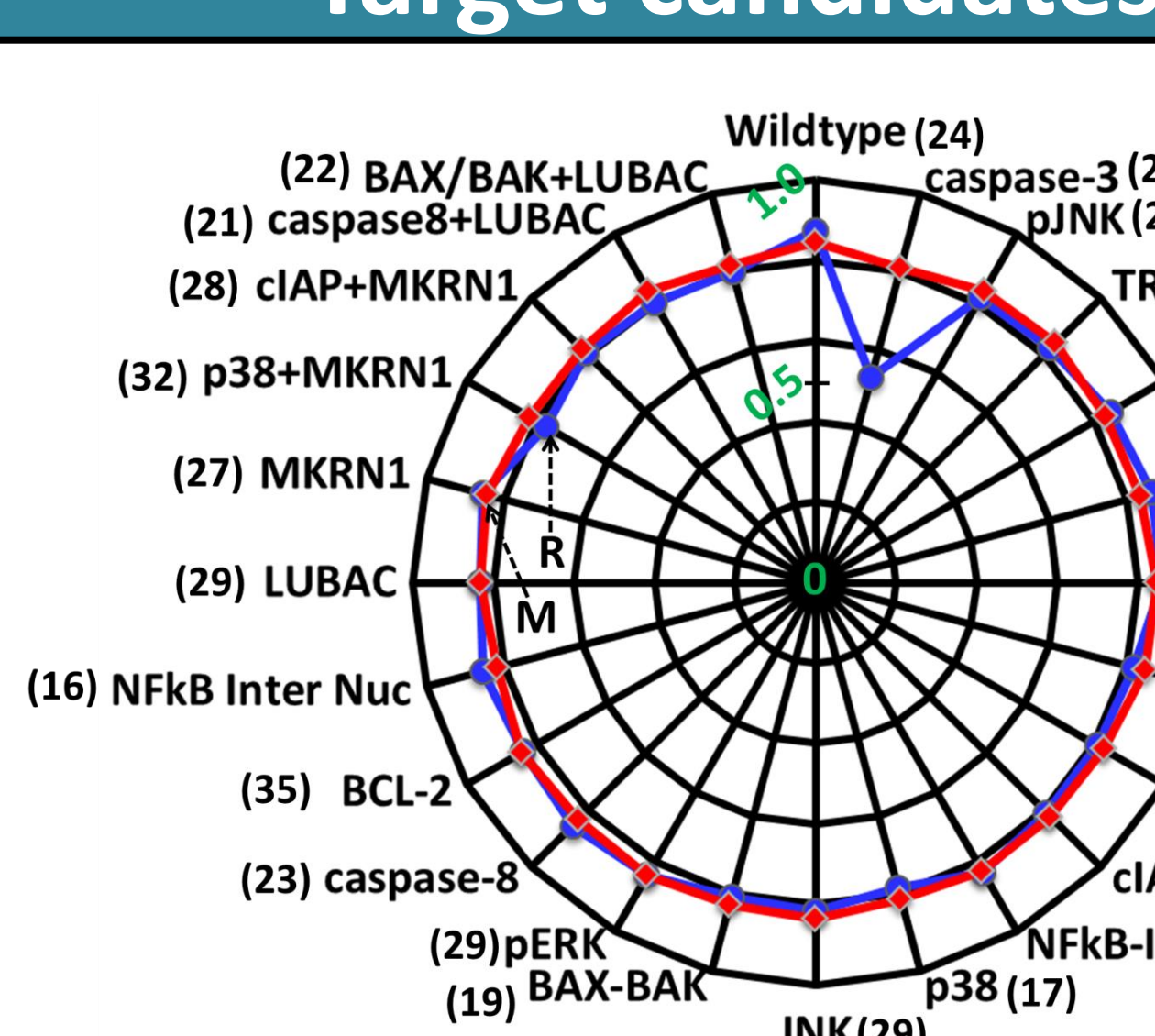


Figure 5: Single or double knock-outs causing reduction in both network modularity (M) and robustness (R). Number next to a candidate within bracket is the number of modules obtained after knock-out.

- ### Conclusions
- Well annotated TNF $\alpha$  signaling network
  - Network predicts that apoptosis ( $P^{WT} = 0.65$ ) is more likely
  - Novel approach for target identification
  - cIAP1/2 & MKRN1: Combination target for switching proliferation to apoptosis phenotype.
  - BAX-BAK & LUBAC: Combination target for switching apoptosis to proliferation phenotype.

- ### References
- Huang *et al.* (2008), Nature Prot.
  - Newman (2011), Nature Physics
  - Stelling *et al.* (2004), Cell
  - Davidich *et al.* (2013), PLoS One
  - Saadatpour and Albert (2013), Methods
- ### Acknowledgements
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## Modularity and Robustness

**Modularity (M)** is the fraction of weighted directed edges between nodes in a module vs expected fraction if randomly connected.<sup>2</sup> Range for M is [0,1]  
**Robustness (R)** is the persistence of the network's characteristic behavior under perturbations.<sup>3</sup> Range for R is [0, 1]