

University of Pavia Dep. of Electrical, Computer and Biomedical Engineering

PREDICTION AND SIMULATION OF MULTI-TARGET THERAPIES FOR TRIPLE NEGATIVE BREAST CANCER THROUGH A NETWORK-BASED DATA INTEGRATION APPROACH

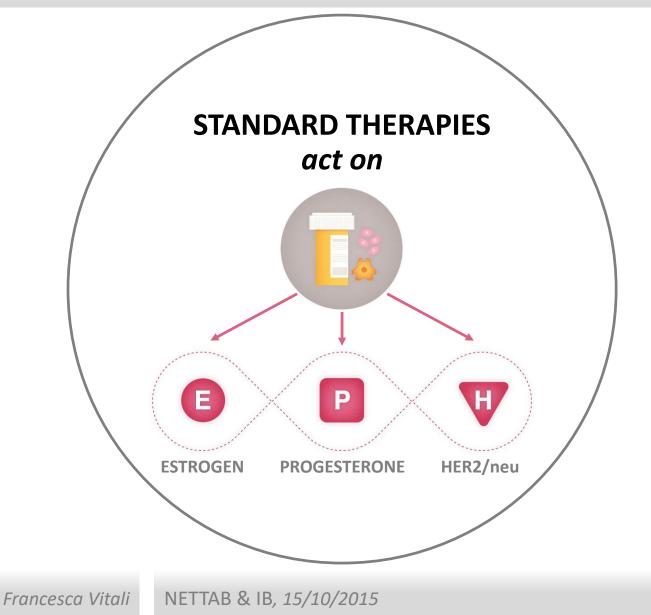
F. Vitali, L.D. Cohen, A. Amato, A. Zambelli, R. Bellazzi





NETTAB & Integrative Bioinformatics 2015, 15/10/15

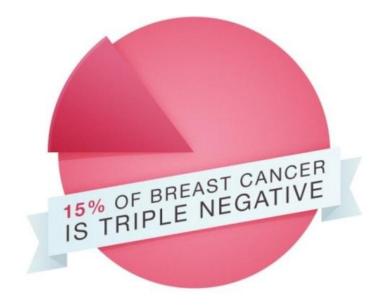
Breast Cancer THERAPIES



mathematical modeling

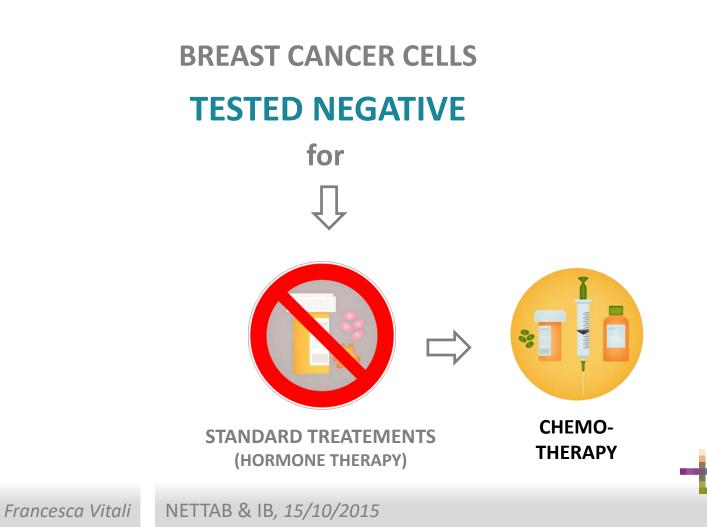
and synthetic biology

Triple Negative Breast Cancer - TNBC





Triple Negative Breast Cancer - TNBC



mathematical modeling

and synthetic biology



NETWORK BASED APPROACH

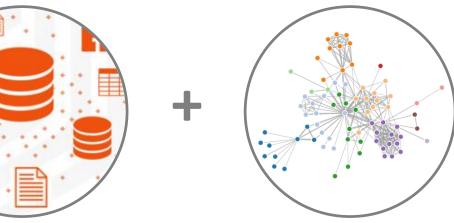
SEARCH FOR NEW THERAPIES



BY EXPLOITING THE PRIOR KNOWLEDGE

DATA SOURCES

INTERACTION NETWORKS

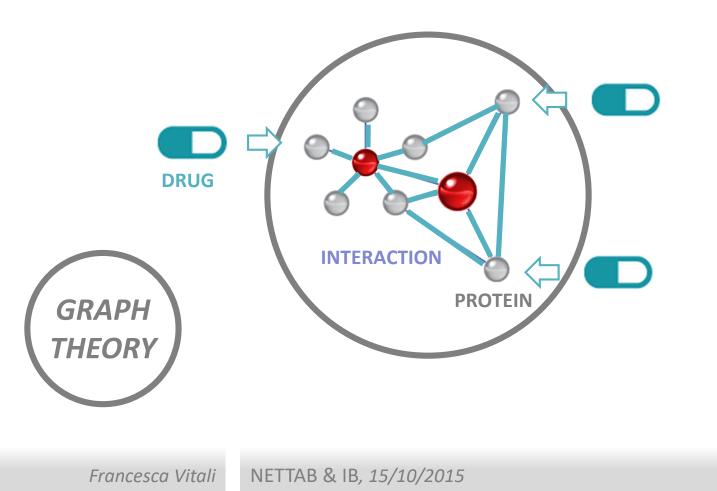






SYNERGISTIC APPROACH

PROTEIN–PROTEIN INTERACTION (PPI) NETWORK





PIPELINE





Francesca Vitali

NETTAB & IB, 15/10/2015



PIPELINE



Selection of CANDIDATE TARGET COMBINATIONS

through TOPOLOGICAL SCORE OF DRUG SYNERGY (TSDS)



TSDS APPROACH

DISEASE PROTEINS (e.g. mj ated or differentially nes) expr 🌸 STRING PPI repository W Experimental evidence **PPI NETWORK**

EDGE WEIGHT W

= LEVEL OF CONFIDENCE





CONSTRAINTS: TOPOLOGICAL FEATURES and DRUGGABILITY











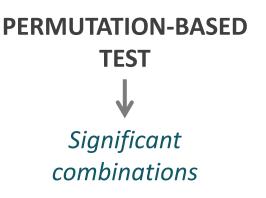
of drug effect



TOPOLOGICAL SCORE OF DRUG SYNERGY

MULTI-TARGET RANKING based on DPs RECHABILITY

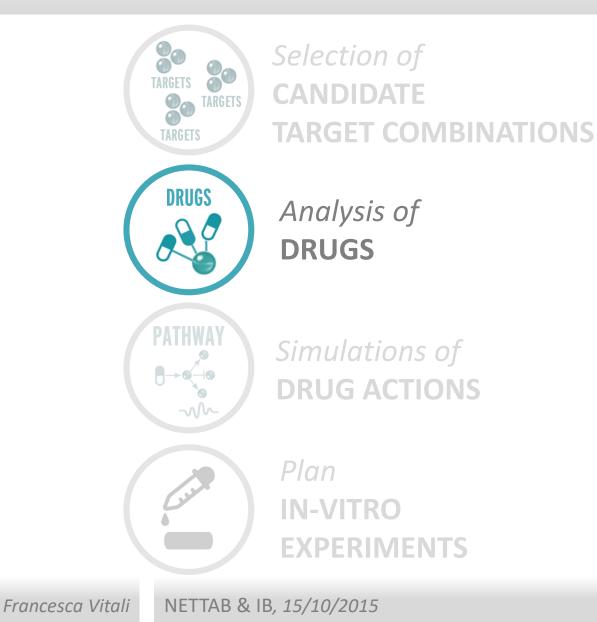




F. Vitali et al. , JBI, 2013

BMS bioinformatics mathematical modeling and synthetic biology

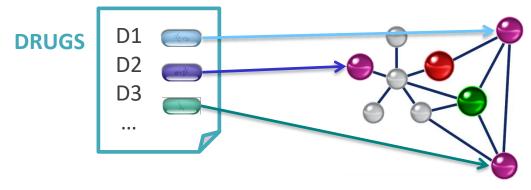
PIPELINE



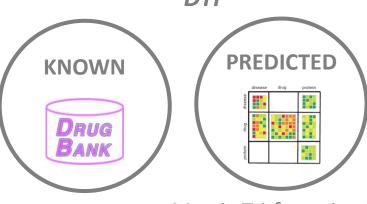




DRUG ANALYSIS







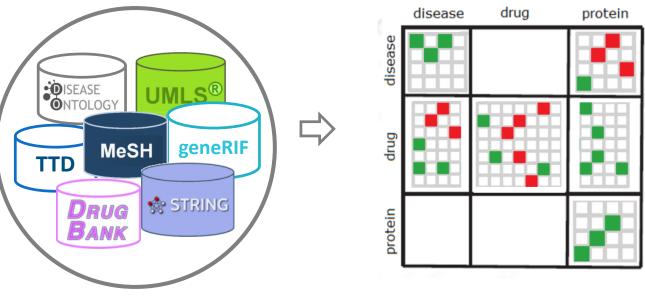
Matrix Tri-factorization





MATRIX TRI-FACTORIZATION





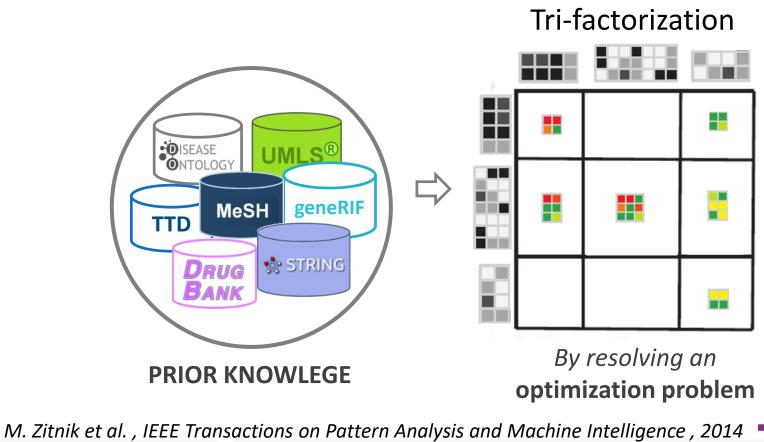
PRIOR KNOWLEGE

M. Zitnik et al., IEEE Transactions on Pattern Analysis and Machine Intelligence, 2014

BMS bioinformatics mathematical modeling and synthetic biology



MATRIX TRI-FACTORIZATION



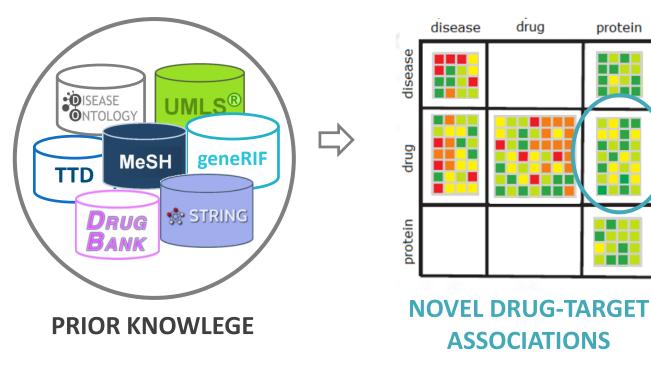




MATRIX TRI-FACTORIZATION

DTI Prediction

protein

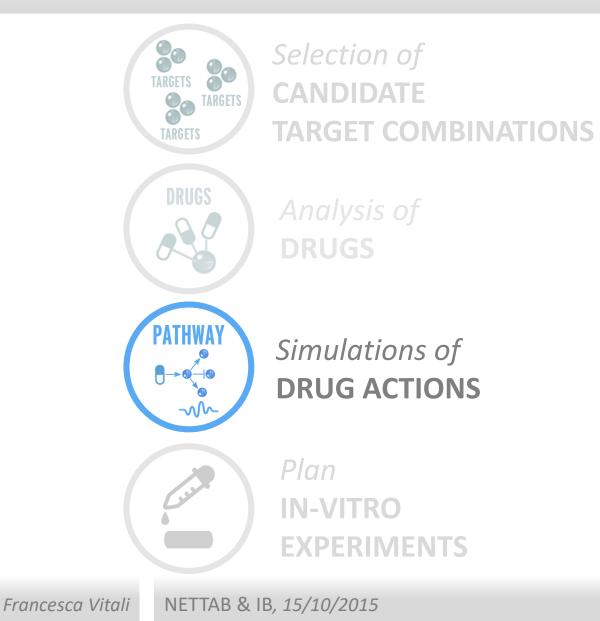


M. Zitnik et al., IEEE Transactions on Pattern Analysis and Machine Intelligence, 2014

mathematical modeling and synthetic biology

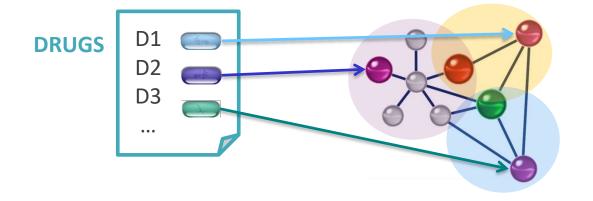
NETTAB & IB, 15/10/2015 Francesca Vitali

PIPELINE







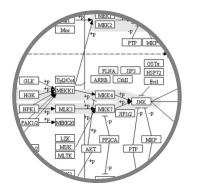


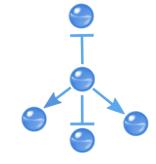






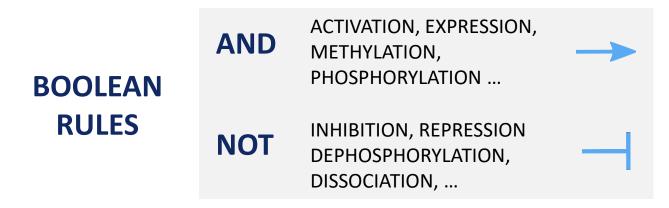
PATHWAY MODELS





KEGG PATHWAY

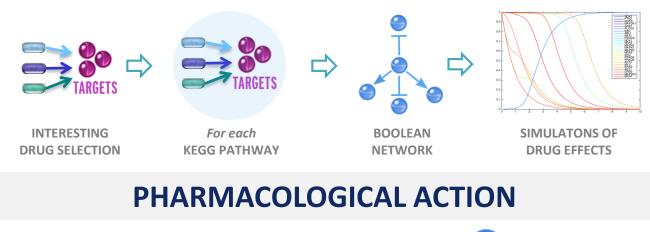
BOOLEAN NETWORK

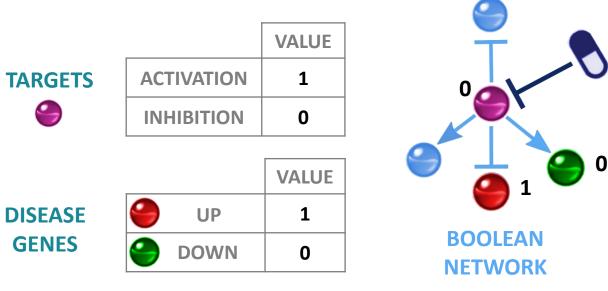






PHARMACOLOGICAL ACTION MODEL









DRUG SIMULATIONS

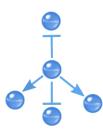


DRUG ADMINISTRATION SIMULATIONS

BOOLEAN MODEL

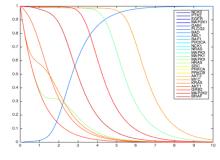
CONTINUOUS MODEL

Logical expressions





System of ODEs



NODE BEHAVIORS

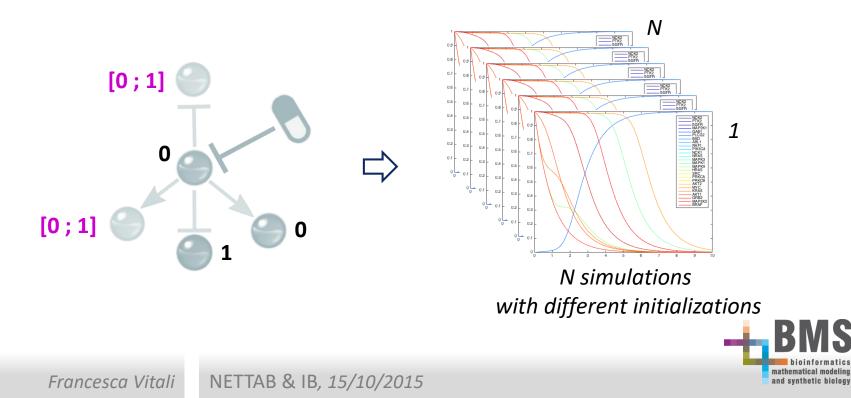
Krumsiek J. et al., BMC bioinformatics, 2010





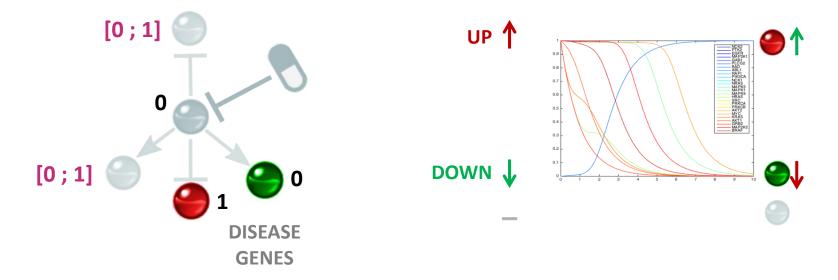


MONTECARLO SIMULATIONS





RESULT ANALYSIS



DRUG/DRUG COMBINATION

has to **REGULARIZE** as much as possible **Disease Genes** *while* **other nodes** must **NOT be PERTURBED**

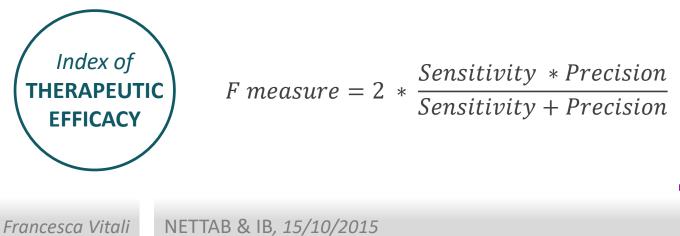


RESULT ANALYSIS

CONFUSION MATRIX

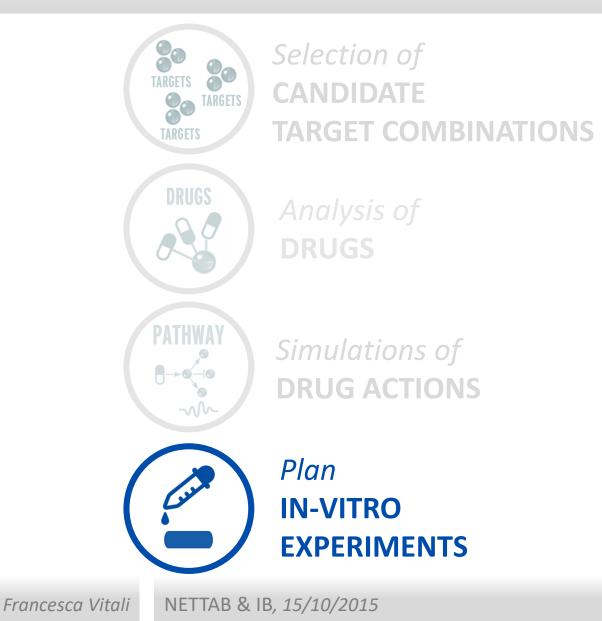
	Disease Genes	Disease Genes		
Regularized Genes	ТР	FP	Tot RG	$Sensitivity = \frac{11}{Tot DG}$
Regularized Genes	FN	FN	Tot RG	$\int Precision = \frac{TP}{Tot PG}$
	Tot DG	Tot \overline{DG}	N	

AIM: Maximization of regularized disease genes (TP)





PIPELINE



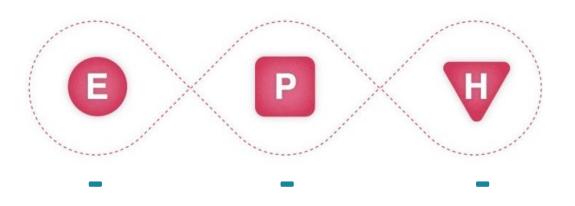






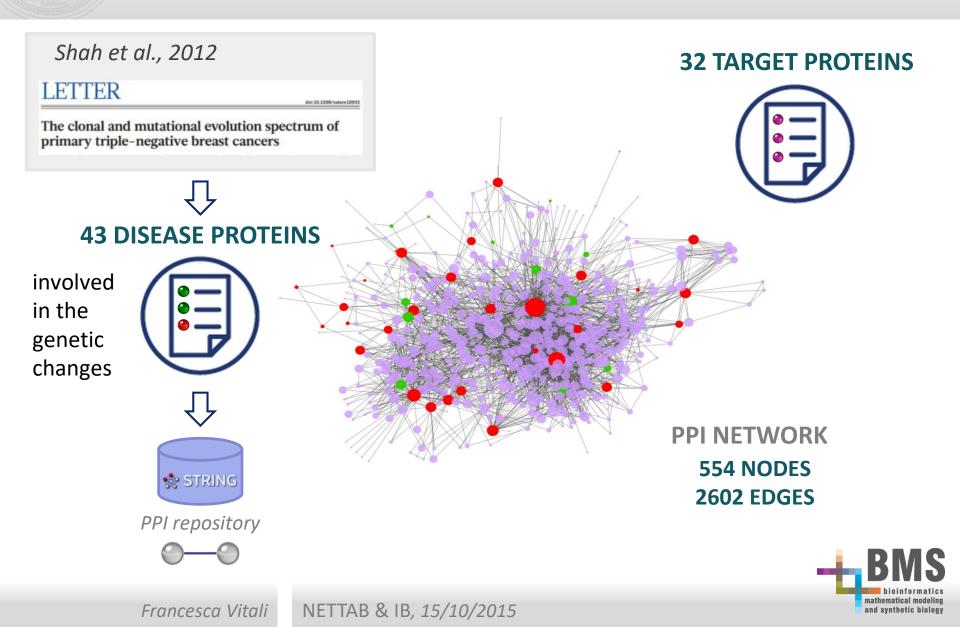
BREAST CANCER CELLS TESTED NEGATIVE

for

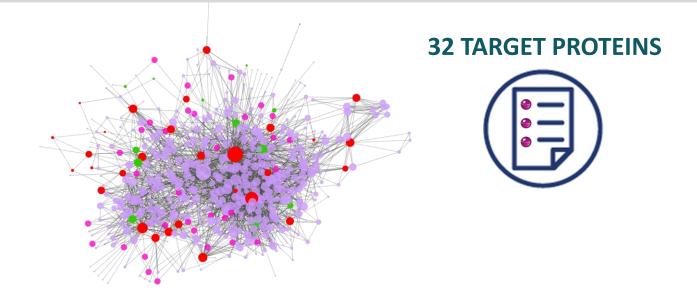




RESULTS - TSDS APPROACH



RESULTS - TSDS APPROACH



TOPOLOGICAL SCORE OF DRUG SYNERGY

MULTI-TARGET RANKING



134 combinations of16 different proteins

None are currently used in ongoing clinical trials



RESULTS - DRUG AND PATHWAY ANALYSIS

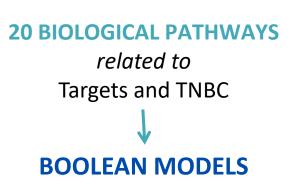


IMATINIB PATHWAY ANALYSIS

...



MAPK signaling pathway **Jak-STAT** signaling **Toll-like receptor signaling ErbB** signaling Phosphatidylinositol signaling





RESULTS - DRUG AND PATHWAY ANALYSIS



IMATINIB PATHWAY ANALYSIS



ELECTIVE DRUG FOR

- Chronic myeloid leukemia
- Gastro-intestinal stromal tumors

REPOSITIONING?

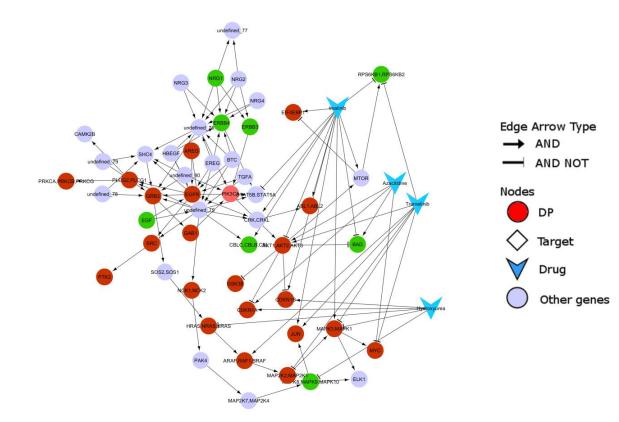


0.689500 0.683210 0.680323 0.679607 0.678105 0.677471 0.677439 0.676923 0.676923 0.670303 0.668500 0.666621



RESULTS - PATHWAY ANALYSIS

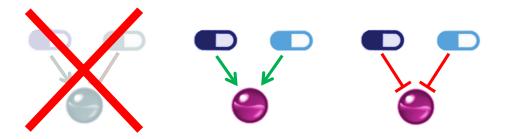
MAPK signaling pathway – BOOLEN NETWORK





RESULTS - PATHWAY ANALYSIS

POLYPHARMACOLOGY = multiple drugs acting on different targets



NO OPPOSITE DRUG-TARGET INTERACTIONS

DRUG ADMINISTRATION SIMULATIONS

- 1. Imatinib
- 2. Imatinib + Vemurafenib
- 3. Imatinib + Flucytosine



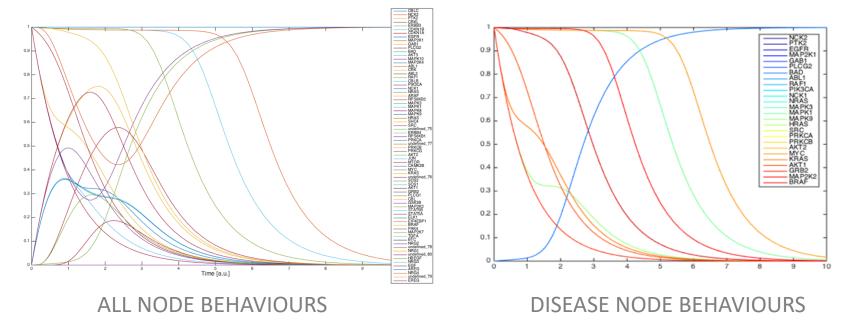
4. Imatinib + Vemurafenib + Flucytosine



RESULTS – MONTECARLO SIMULATIONS

for each DRUG: for each PATHWAY: MONTECARLO SIMULATIONS by applying Odefy

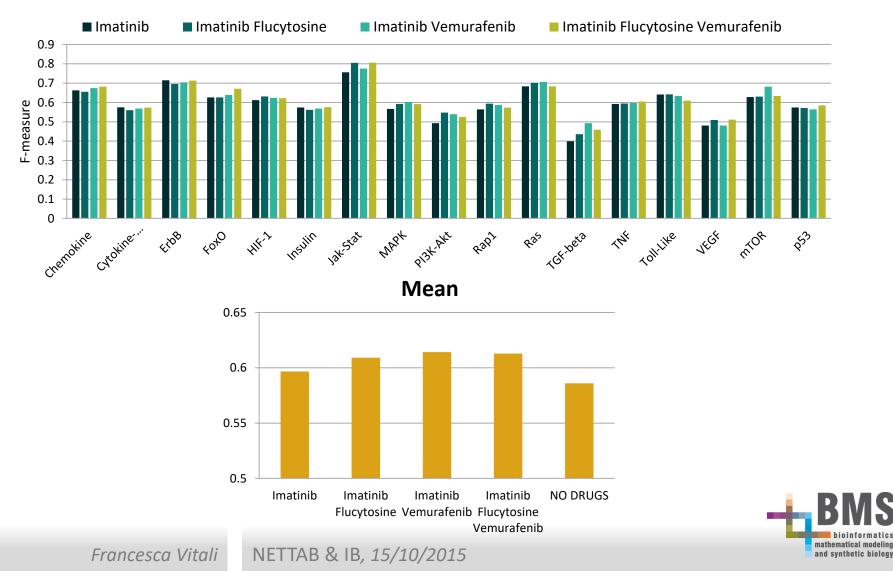
Example: Imatinib – MAPK pathway





RESULTS – MONTECARLO SIMULATIONS

F-measures in pathways

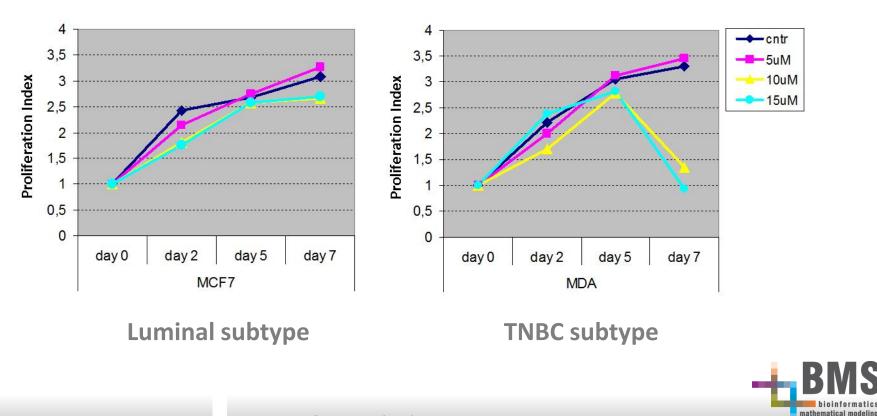


IN VITRO RESULTS



MTT assay – Evaluation of cell viability

2-5-7 days of treatment with different concentration of Imatinib



and synthetic biology

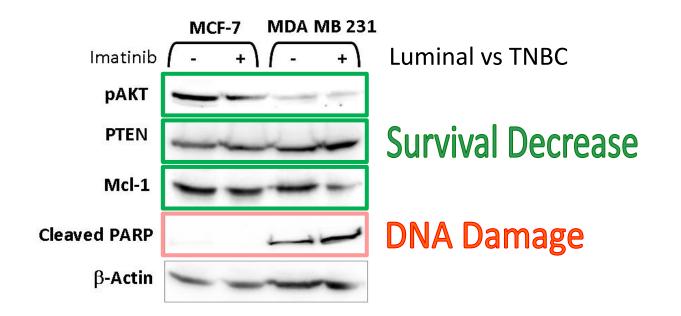


IN VITRO RESULTS



Western Blot assay – Disease gene evaluation

One week treatment with 10 μ Imatinib



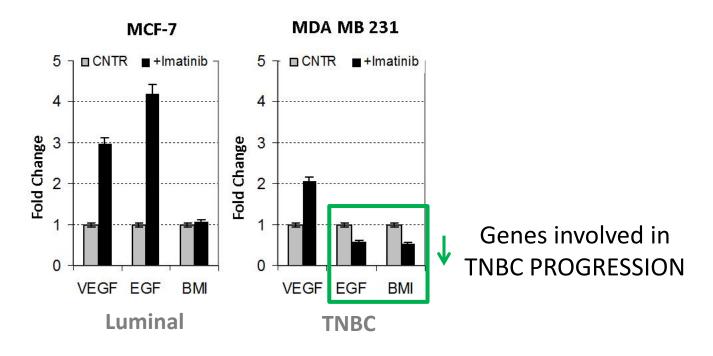


IN VITRO RESULTS



Gene expression assay – Disease gene evaluation

One week treatment with 10μ Imatinib









IN-VITRO EXPERIMENTS ON GOING of drug combinations





- The application to breast cancer
 - > highlights potential targets and drugs
 - enables the selection of top candidate pathways and potential drug combinations in a variety of multifactorial diseases
- The developed approach can easily be applied to other complex diseases
 - Network construction based on a list of mutated proteins in patient's specific proteomic or genomic background – precision medicine





- Combinations of different drugs
- Future works:
 - Bayesian networks instead of Boolean graphs

