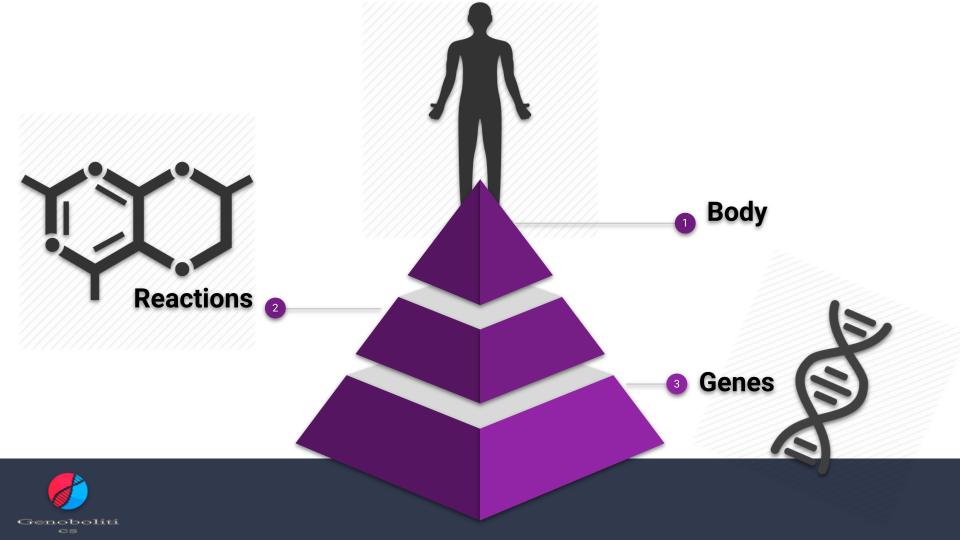


Abdurrahman Aboudakila Ammar Raşid

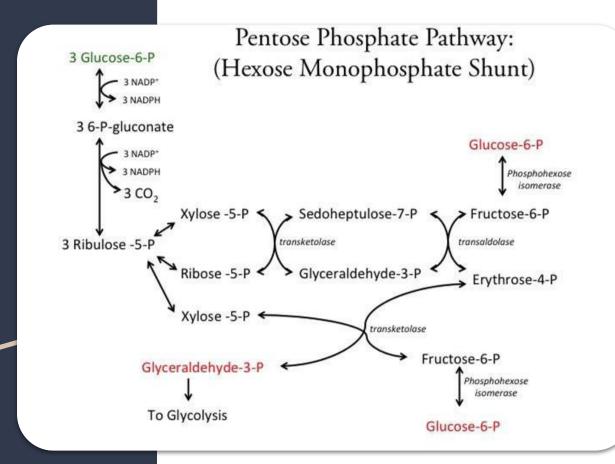
Outline



- 1. Introduction
- 2. Problem definition
- 3. Novelty and Methods
- 4. Results and Discussion



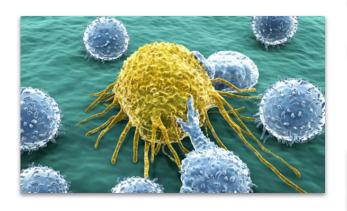
Pathways

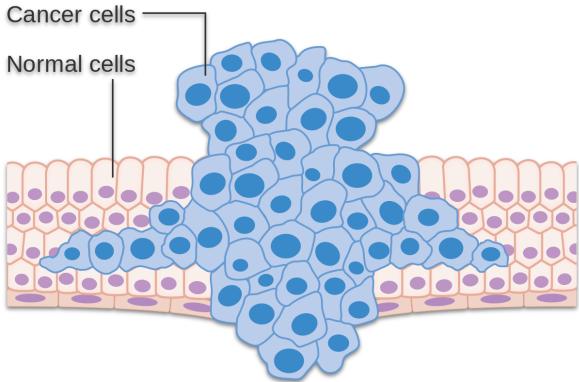






Cancer







Problem Definition

Inputs

Gene Expression Arrays of healthy and patient samples of various diseases

	TCGA-A8-A0A7-01	TCGA-A8-A07G-01	TCGA-A8-A08R-01
15E1.2	-0.697667	-1.588	-1.64925
2'-PDE	0.100687	-0.166	0.746375
7A5	1.3415	-0.01	-0.8525
A1BG	1.711	0.146667	-0.560667
A2BP1	-1.58833	-1.20367	-1.19967
A2M	1.2035	0.9535	0.81
A2ML1	1.5145	0.1755	0.734
A3GALT2	-0.04275	0.456	0.42675
A4GALT	0.5285	0.501667	1.11383
A4GNT	0.5755	-0.1155	0.3605

Goal

- Train a disease classifier
- Detect significant pathways in diseases
- Construct disease-ontology

Methods

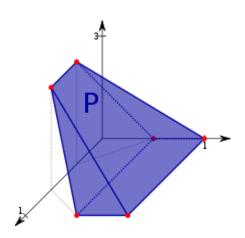
- Dynamic Linear Programming
 - Flux Variability Analysis (FVA)
- Computing Diff values (Features)
- Machine Learning





Dynamic Linear Programming

- Objective Coefficients = Fold Changes
- 2. Solve the objective function = C
- 3. Constrain ObjFun with C
- 4. Repeat for every reaction
- Max+Min Reactions FCs



Objective Function =
$$\sum_{m \in M} \sum_{R \in m_R} m_{fc} * R[m] / m_{TS}$$



Compute Diff Values

- 1. U_{ref} , $L_{ref} = avg(U_i)$, $avg(L_i)$ for i in healthy samples.
- 2. Diff = $((U U_{ref}) + (L L_{ref}))/2$



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Disease Classification

1. Genobolitics Pipeline

a. Nested Cross-Validation

b. 10-Folds

c. 10-Trials

Dimensionality	PCA
Reduction	21 components
Classifier	SVC C = 4.641



Breast-Cancer Case Study

UCSC Xena Agilent G4502A

64 Healthy

534 PT & Metastatic

	test_accuracy	test_f1	test_precisio	test_recall
mean	0.907263	0.947750	0.952113	0.944375
std	0.005368	0.003069	0.002683	0.004738
min	0.901254	0.944038	0.948457	0.936373
25%	0.902843	0.945329	0.950382	0.941929
50%	0.906241	0.947312	0.951827	0.944654
75%	0.912138	0.950291	0.953689	0.945790
max	0.914785	0.952230	0.957194	0.953075

Using Diff Values



Breast-Cancer Case Study

	est_accurac	test_f1	test_precision	test_recall
mean	0.898602	0.945778	0.907145	0.988763
std	0.002527	0.001165	0.018995	0.022036
min	0.894612	0.944174	0.897342	0.949441
25%	0.897947	0.945066	0.899052	0.996261
50%	0.899558	0.945941	0.899106	0.998113
75%	0.899586	0.946848	0.899130	1.000000
max	0.901308	0.946862	0.941097	1.000000





Most significant pathways (based on Raw-values):

'Nucleotide interconversion',
'Purine synthesis',
'Pyrimidine synthesis',
'Vitamin A metabolism'



A lung cancer dataset of **107** samples, consisting of **49** healthy, **58** unhealthy samples was used.

Same procedure used in previous study was applied here.



	test_accuracy	test_f1	test_precision	test_reca	
mean	0.827717	0.830682	0.884762	0.805333	
std	0.016529	0.014782	0.019649	0.015571	
min	0.802323	0.804992	0.856429	0.776667	
25%	0.817626	0.822792	0.874167	0.795000	
50%	0.827323	0.832343	0.885595	0.808333	
75%	0.834116	0.837756	0.896369	0.815833	
max	0.860707	0.858353	0.914762	0.830000	

Lung-Cancer Case Study (GDS3257)



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f1 score, precision and recall reported across 10-trials of nested cross-validation (Diff. scores)

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f1 score, precision and recall reported across 10-trials of nested cross-validation (Diff. scores)





Most significant pathways:

'Blood group synthesis', 'Cholesterol metabolism', 'Eicosanoid metabolism', 'Fatty acid oxidation',

'Folate metabolism',
'Inositol phosphate metabolism',

'Nucleotide interconversion', 'Phosphatidylinositol phosphate metabolism',

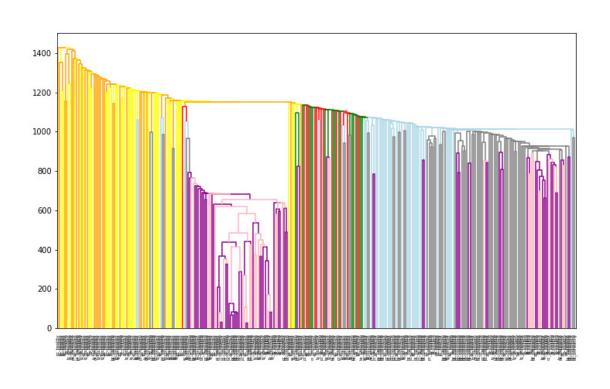
'Pyrimidine synthesis', 'Steroid metabolism',

'Thiamine metabolism', 'Urea cycle'

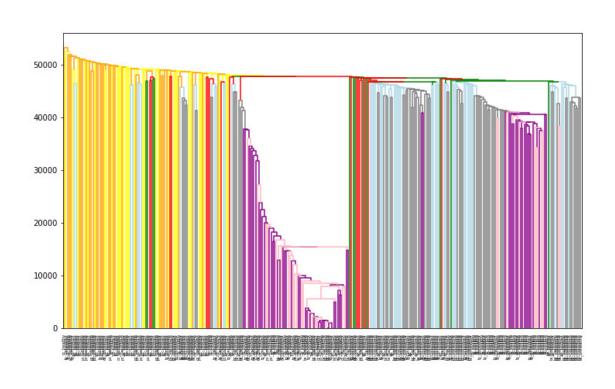
Clustering

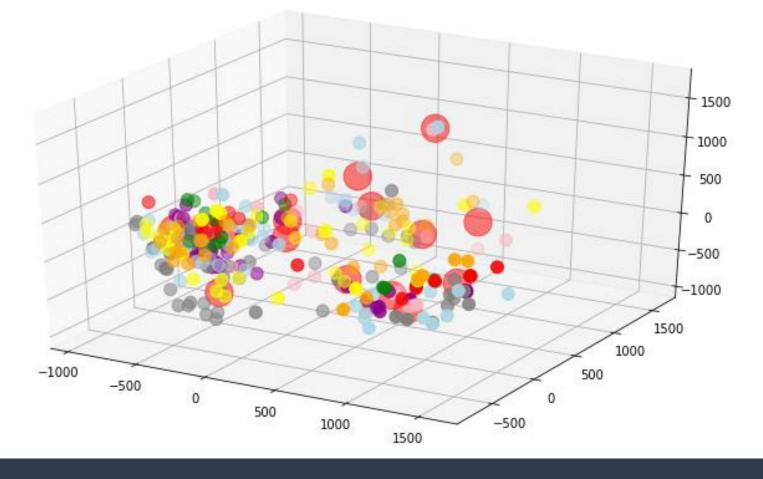
BC healthy	Light Pink
BC Patient	Dark magenta
CC Healthy	Green
CC Patient	Red
LG Healthy	Light Blue
LG Patient	Grey
PC Healthy	Yellow
PC Patient	Orange

Hierarchical Clustering (Diff)

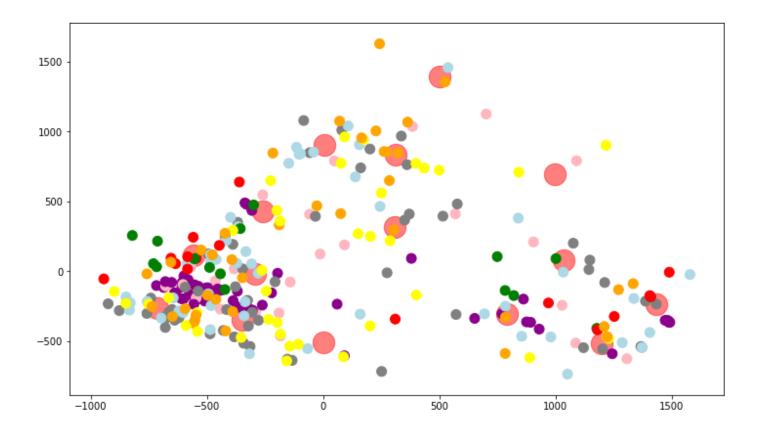


Hierarchical Clustering (Raw)





Clustering (PCA diff. score)



Clustering (PCA diff. score)

	LDA_2d	PCA_2d	k-best_2d	t-SNE_2d	LDA_3d	PCA_3d	k-best_3d	t-SNE_3d
completeness	0.378431	0.146051	0.165582	0.161552	0.433107	0.138710	0.180055	0.095596
homogeneity	0.514187	0.192522	0.222582	0.222492	0.582785	0.175087	0.232406	0.130141
v_measure	0.435986	0.166098	0.189897	0.187187	0.496920	0.154790	0.202908	0.110225

	LDA_2d	PCA_2d	k-best_2d	t-SNE_2d	LDA_3d	PCA_3d	k-best_3d	t-SNE_3d
completeness	0.924294	0.198222	0.332398	0.232236	0.930902	0.138710	0.332398	0.126032
homogeneity	0.991393	0.260313	0.272538	0.316967	0.991870	0.175087	0.272538	0.174126
v_measure	0.956669	0.225064	0.299506	0.268065	0.960420	0.154790	0.299506	0.146226

Clustering Metrics Results

Questions