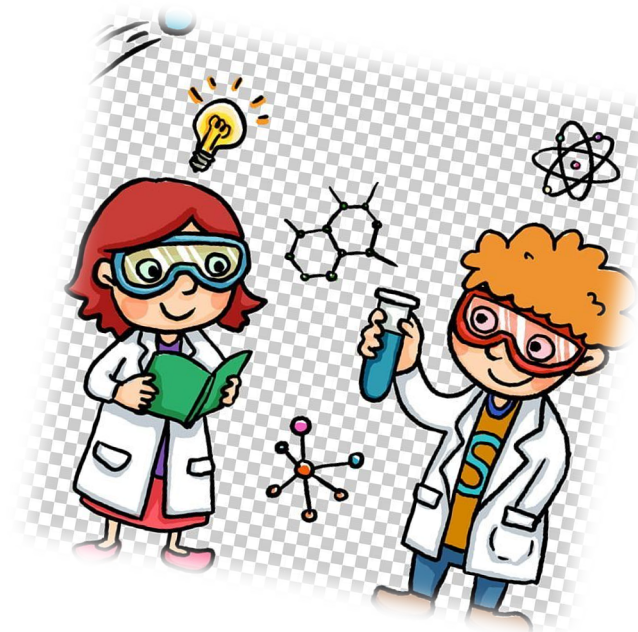


Gene Networks and Drugs

What can we learn by putting
together
bio- and chemo-
informatics?

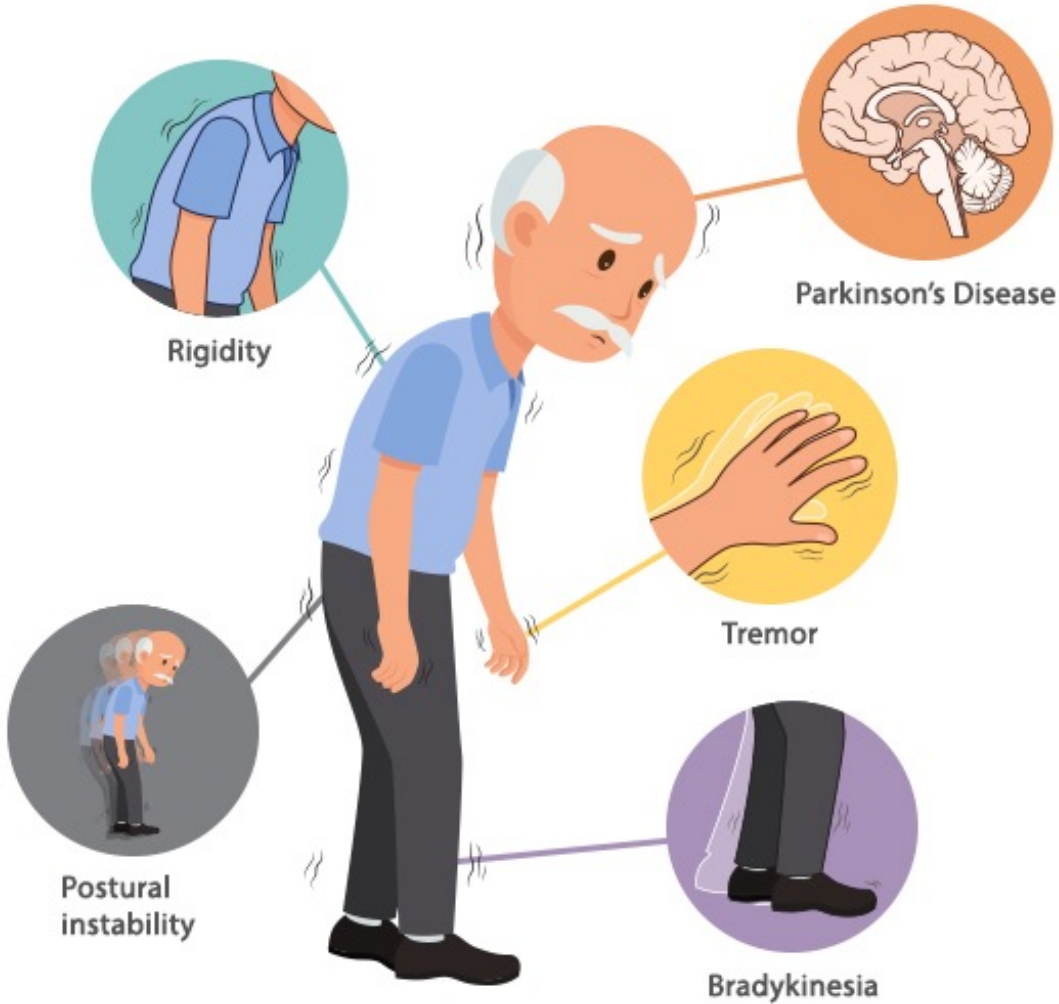


Alexander Kel

geneXplain Germany

Institute of Chemical Biology and Fundamental Medicine, Russia

Parkinson's Disease



Progressive brain disorder that leads to shaking and stiffness

Difficulties with walking, balance and coordination (motor symptoms)

Non-motor symptoms

Degeneration of neurons that synthesise the neurotransmitter dopamine

Norepinephrine-synthesising neurons are also affected

Formation of Lewy bodies in neurons

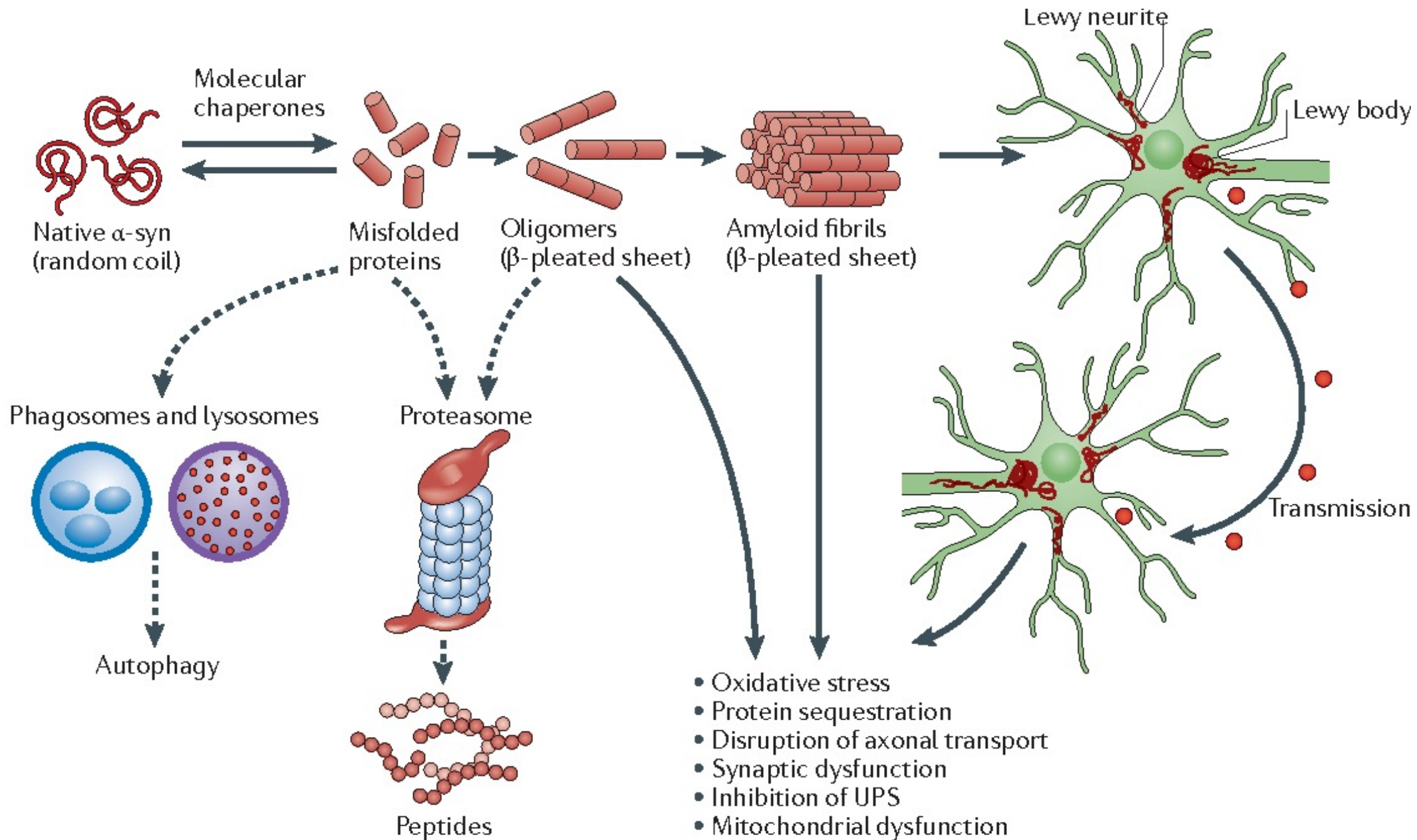
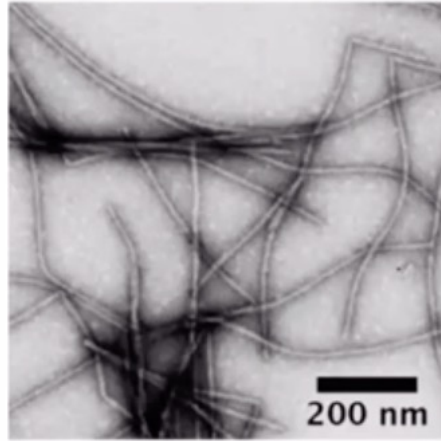
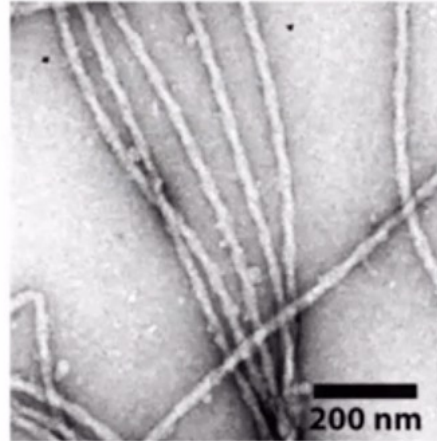


Figure 1 | Hypothetical model of α -syn toxicity and spread of pathology in PD and PDD. Under physiological

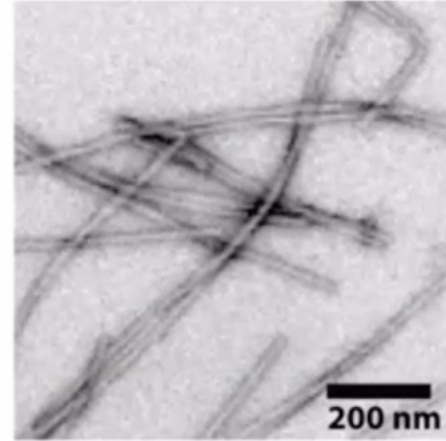
α Syn fibrils come in many forms



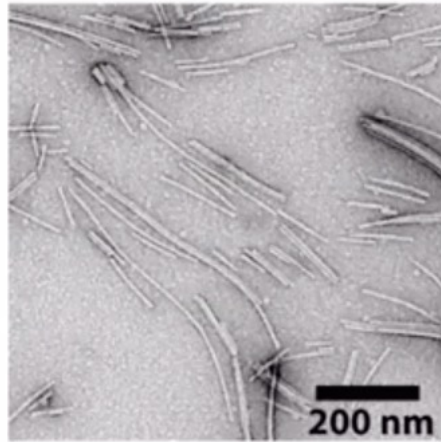
α Syn Fibrils



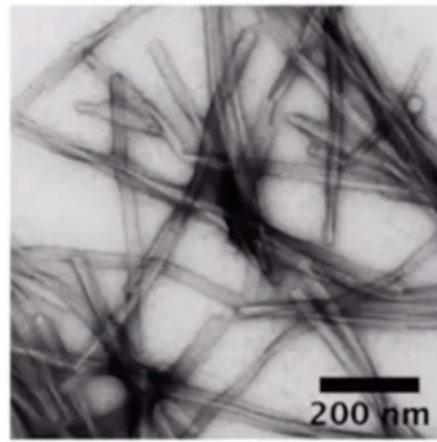
α Syn Fibrils 65



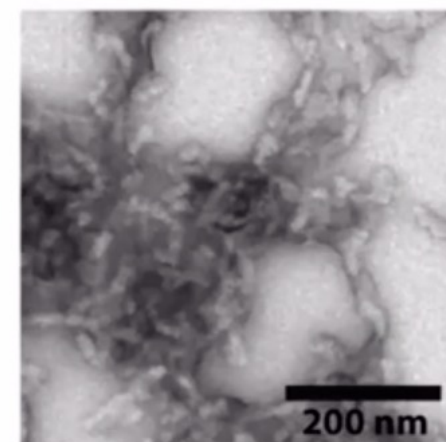
α Syn Fibrils 91



α Syn Fibrils 110



α Syn Ribbons



α Syn assemblies derived from a DLB case

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> [Sci Rep](#). 2018 Nov 1;8(1):16165. doi: 10.1038/s41598-018-34490-9.

The small molecule alpha-synuclein misfolding inhibitor, NPT200-11, produces multiple benefits in an animal model of Parkinson's disease

Diana L Price ¹, Maya A Koike ^{2 3}, Asma Khan ², Wolfgang Wrasidlo ², Edward Rockenstein ⁴, Eliezer Masliah ⁴, Douglas Bonhaus ²

Affiliations + expand

PMID: 30385782 PMCID: [PMC6212487](#) DOI: [10.1038/s41598-018-34490-9](#)

Free PMC article

Abstract

Accumulation of alpha-synuclein (ASYN) in neurons and other CNS cell types may contribute to the underlying pathology of synucleinopathies including Parkinson's disease (PD), dementia with Lewy bodies (DLB) and Multiple Systems Atrophy (MSA). In support of this hypothesis for PD, ASYN

FULL TEXT LINKS



ACTIONS

“ Cite

☆ Favorites

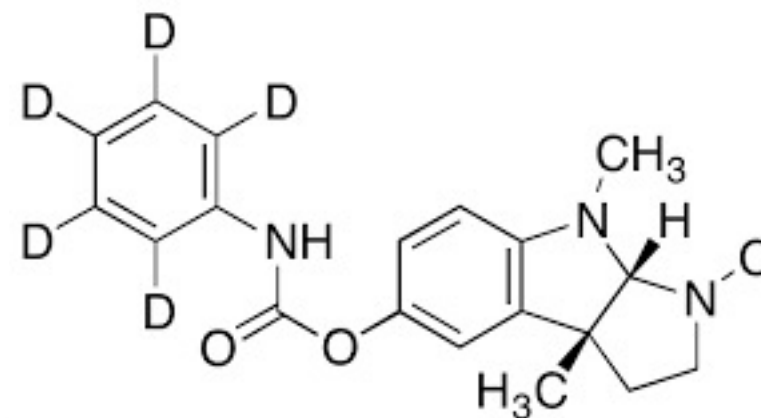
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PAGE NAVIGATION

< Title & authors

What were the results of the clinical study?



In 2019, UCB initiated a US-based multicenter Phase Ib clinical trial of UCB0599 in Parkinson's. 31 people with Parkinson's (Hoehn-Yahr stage 1–3, aged 40–80 years) were recruited and they received two doses of either UCB0599 (n=21) or placebo (n=10) over four weeks ([Click here](#) to read the details of the trials). This was the first time the drug was tested in people with PD.

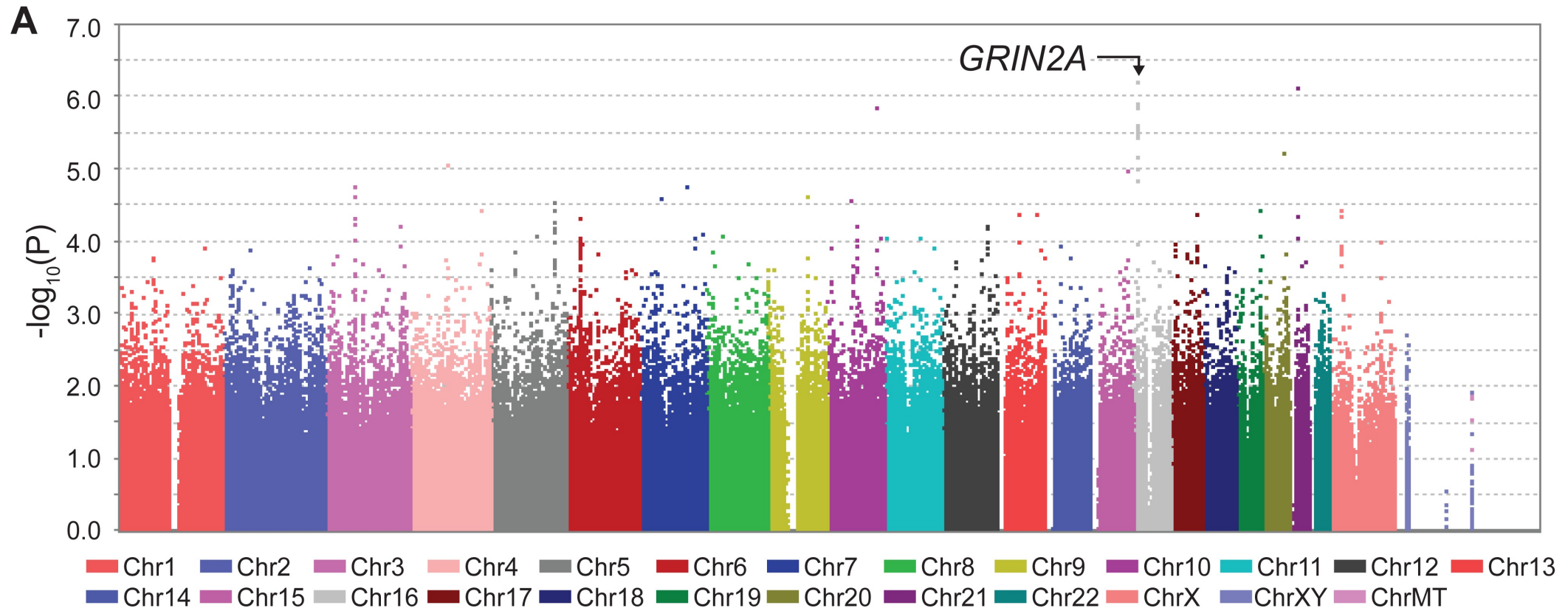
The results of the study indicate that UCB0599 was safe and well tolerated in the study. There was little difference between the UCB0599- and placebo-treated groups in terms of treatment-emergent adverse events (headache being the most common event in both groups).

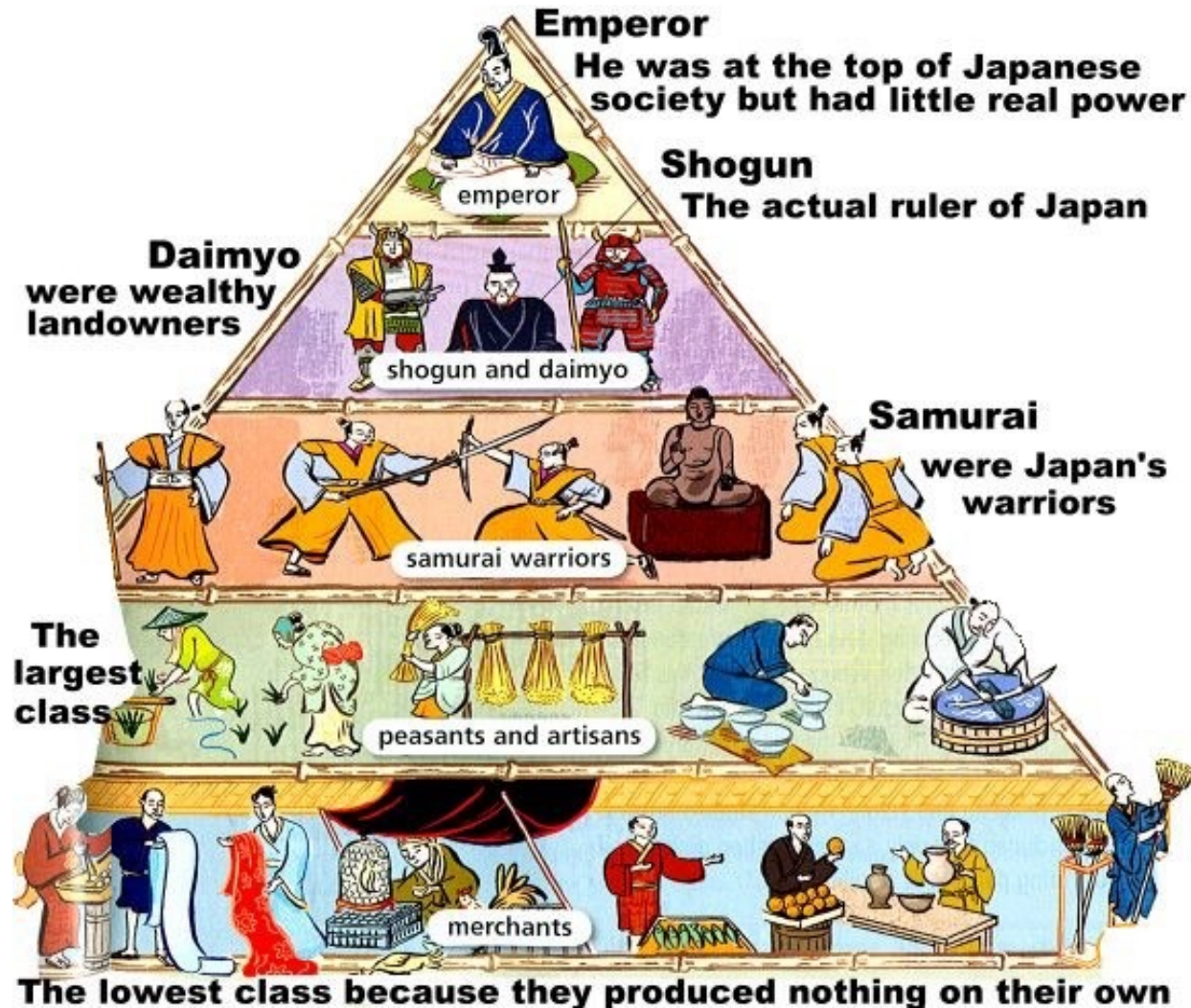
Genome-Wide Gene-Environment Study Identifies Glutamate Receptor Gene *GRIN2A* as a Parkinson's Disease Modifier Gene via Interaction with Coffee

Taye H. Hamza, Honglei Chen, Erin M. Hill-Burns, Shannon L. Rhodes, Jennifer Montimurro, Denise M. Kay, Albert Tenesa, Victoria I. Kusel, Patricia Sheehan, Muthukrishnan Easwarkh Haydeh Payami [view all]

Published: August 18, 2011 • <https://doi.org/10.1371/journal.p>

Epidemiological studies have firmly established that coffee drinking is inversely associated with risk of developing Parkinson disease (PD) [1], and clinical trials of caffeine for treatment of PD have shown symptomatic benefit [2]–[4]. Identifying genes that modulate the efficacy of coffee may therefore have pharmacogenomic potential for prevention and treatment.





➔ Master-regulator

➔ Signal transduction

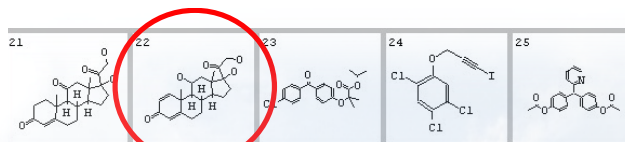
➔ Transcription Factors

➔ miRNAs

➔ Genes of metabolism

➔ Genes of cell cycle, development, communication

We build a pyramid



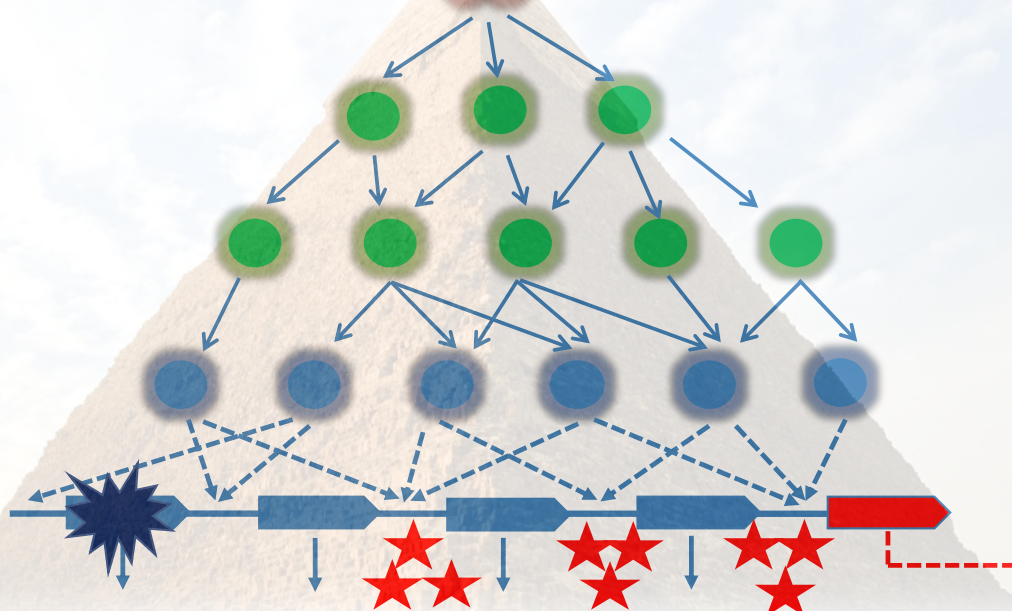
Master-regulator



NGS



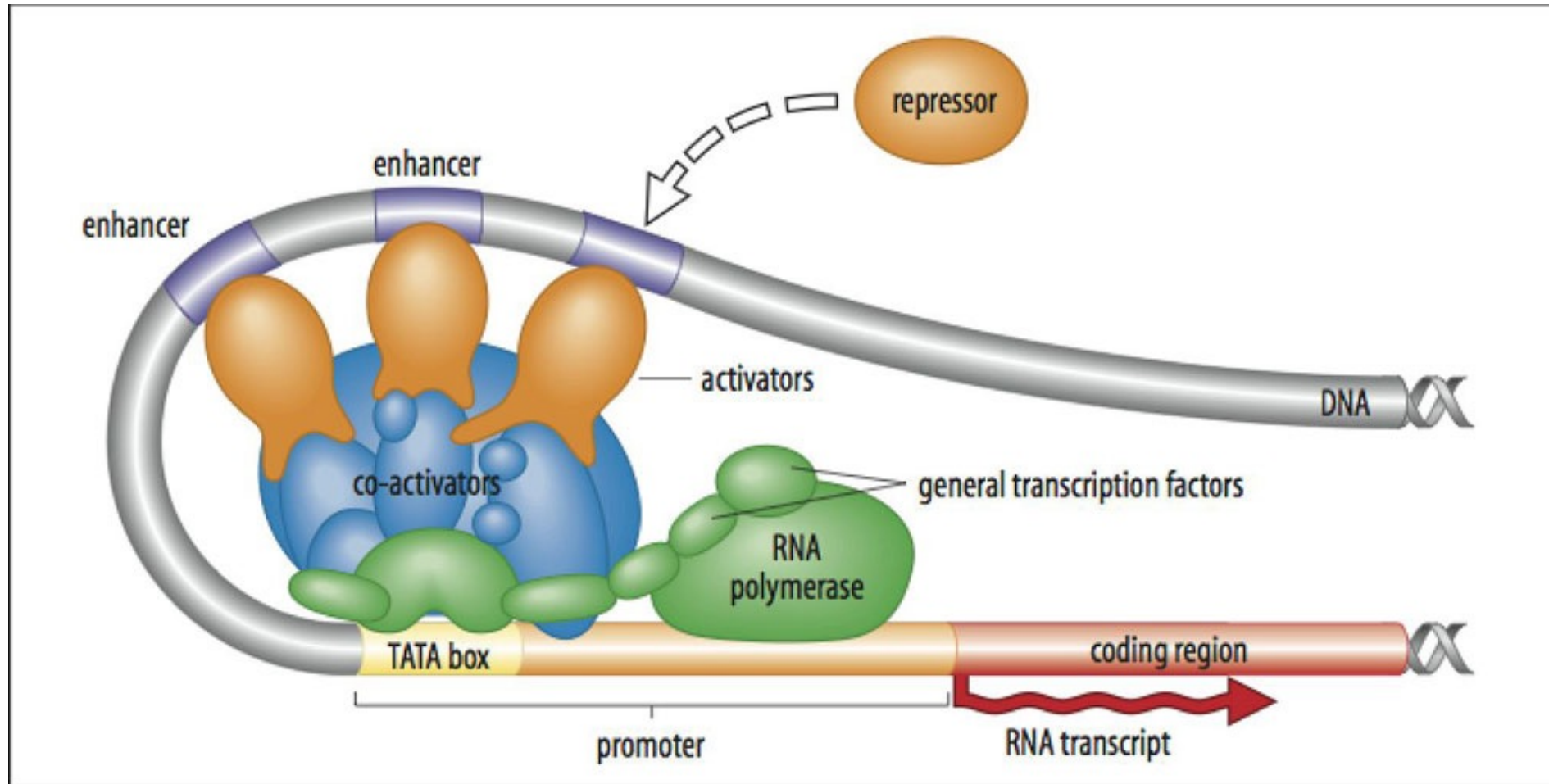
RNA-seq, Exome,
Methylome, CHIP-seq,
Proteome, Metabolome



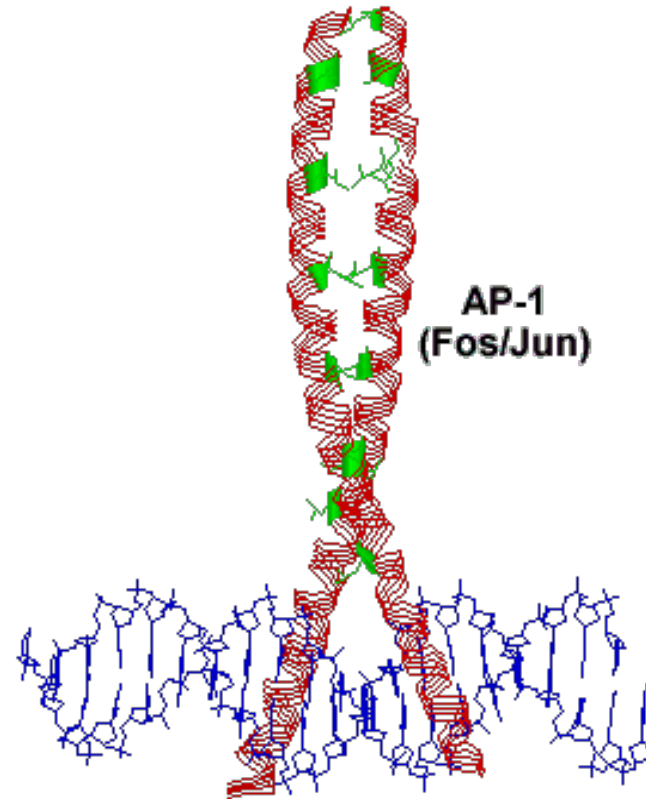
AI



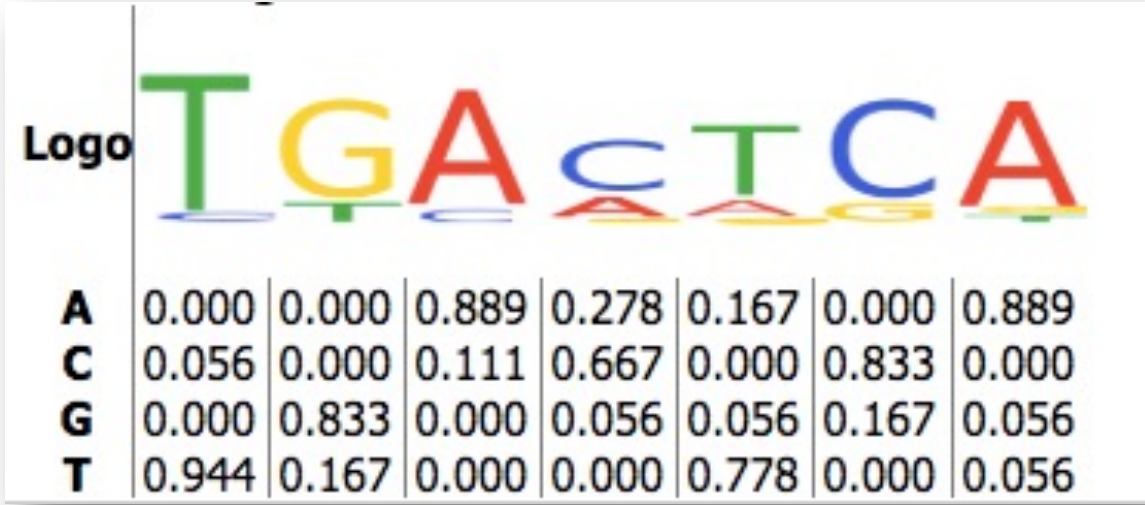
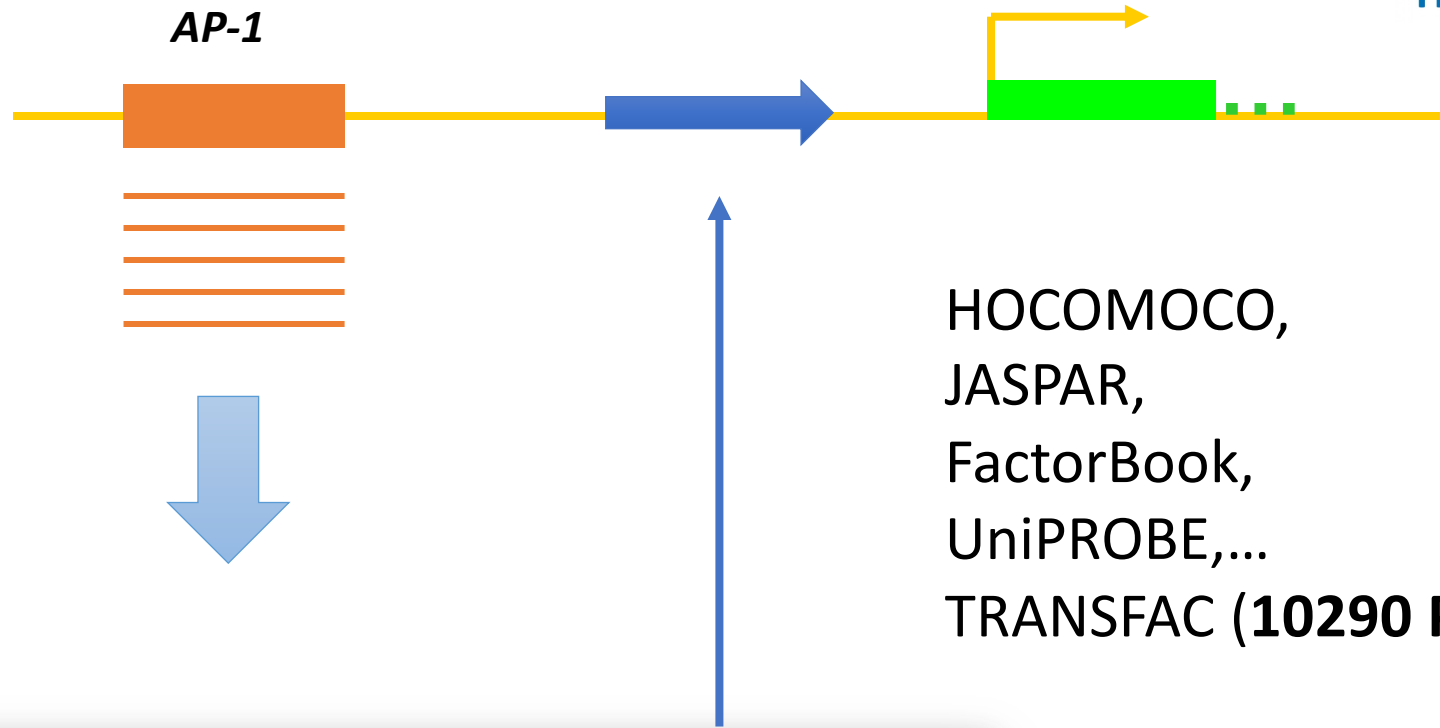
TFs can read between genes



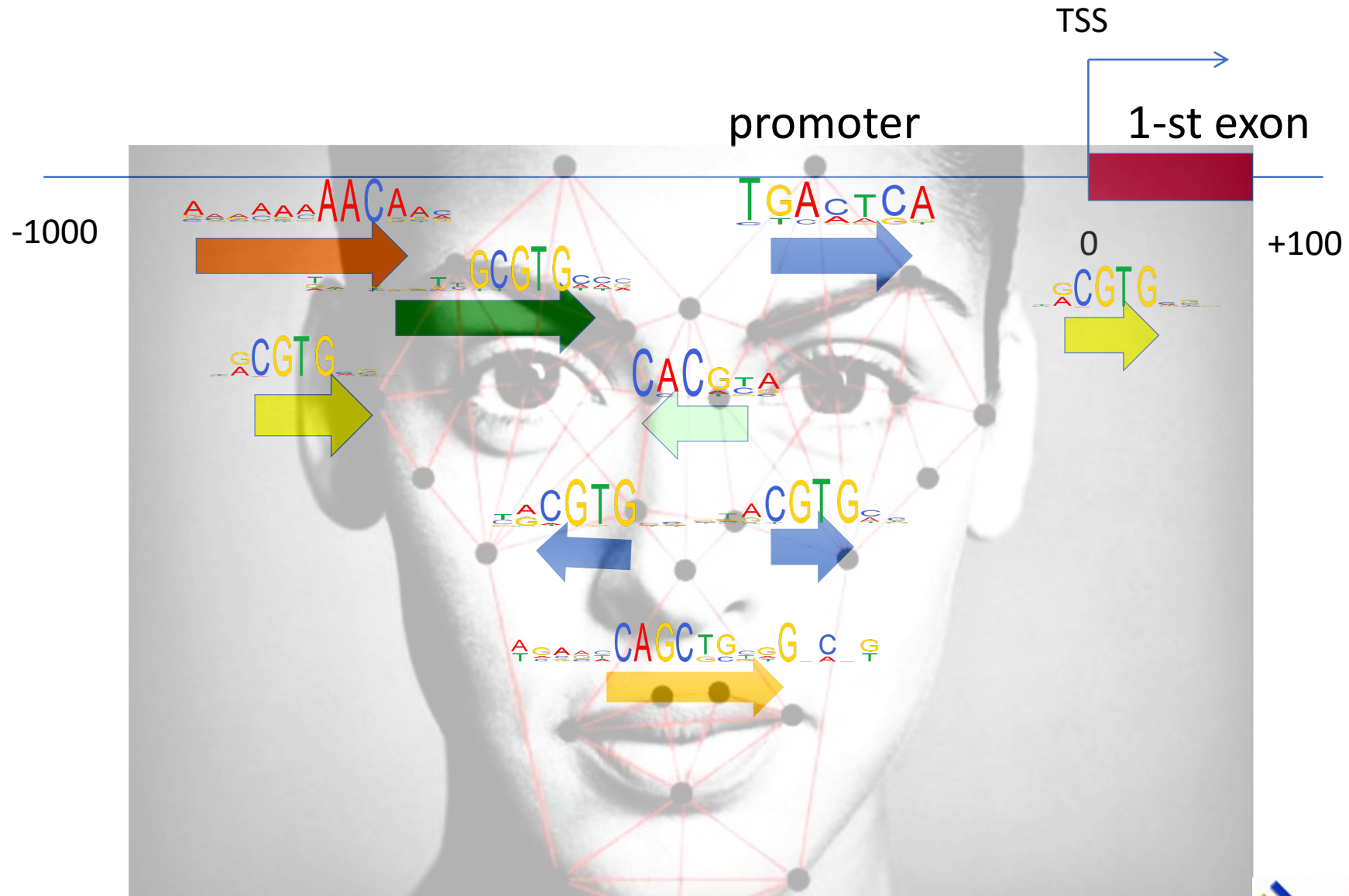
AP-1 (human)



Search for new TF binding sites with PWMs



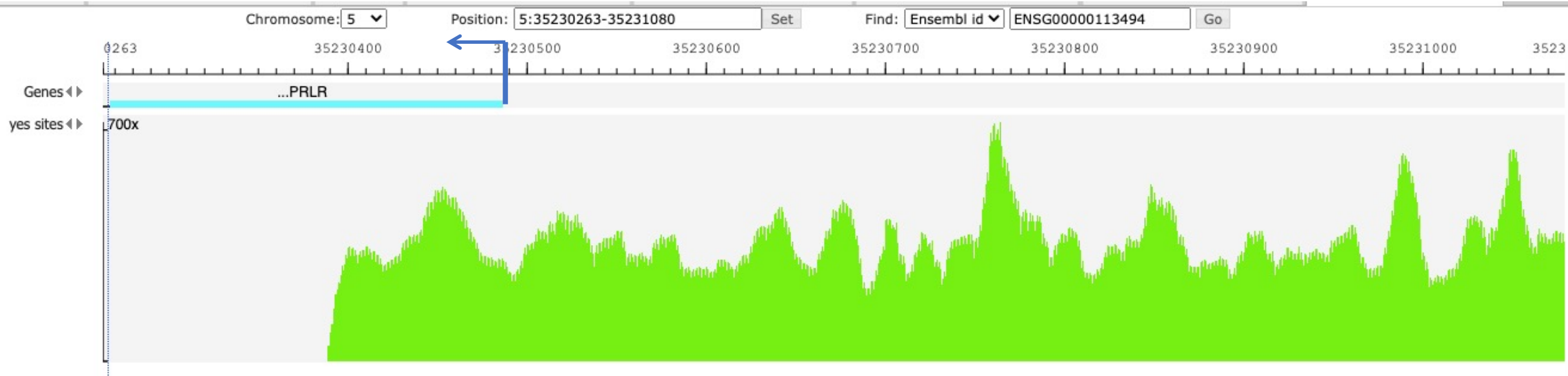
We should try to recognize promoters of active genes in PD



Promoter face



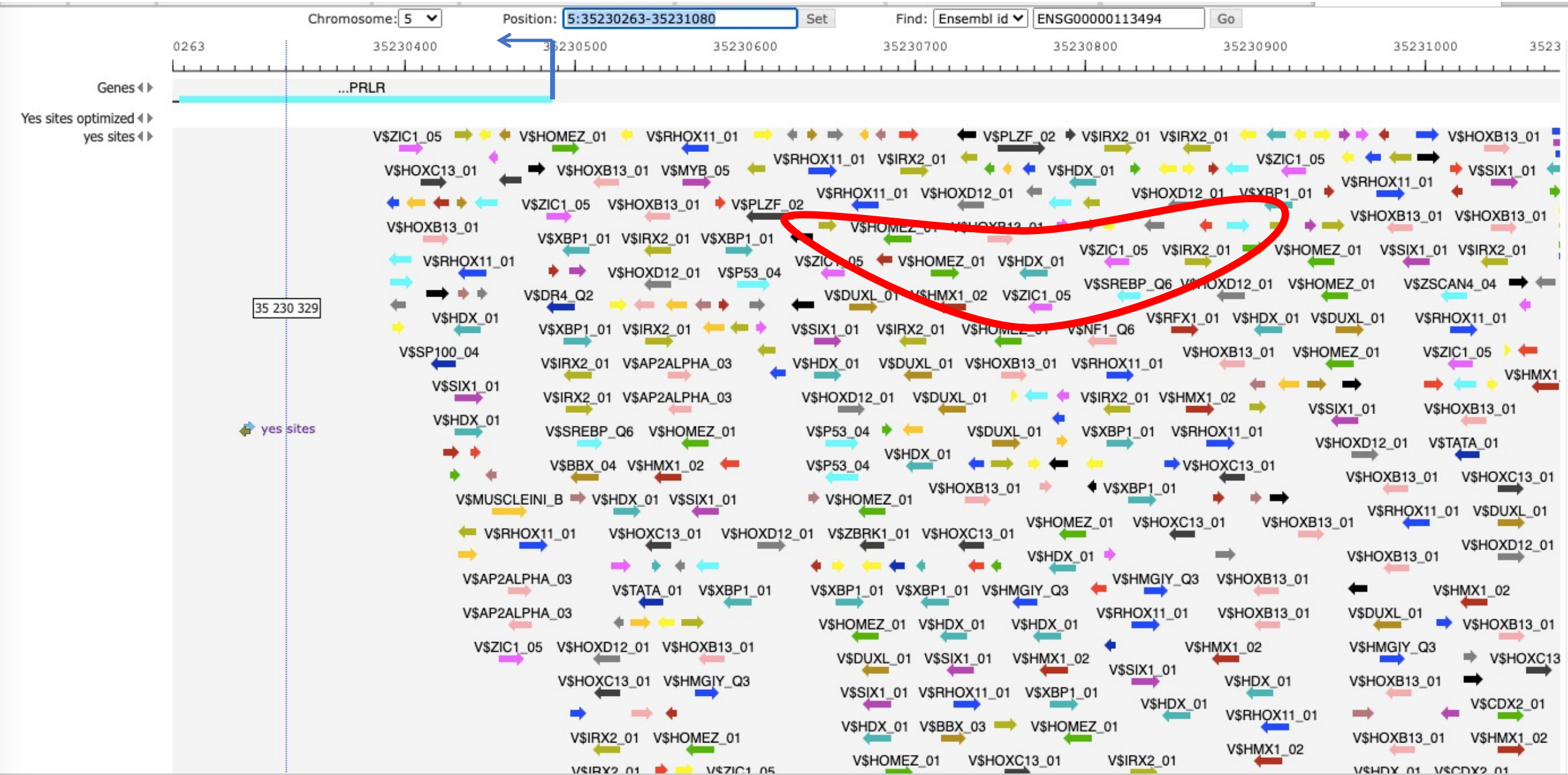
TFBS of 5260 TF motifs



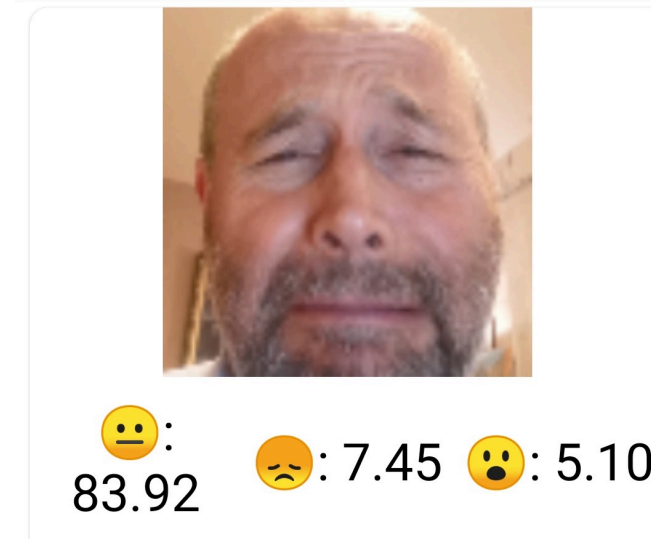
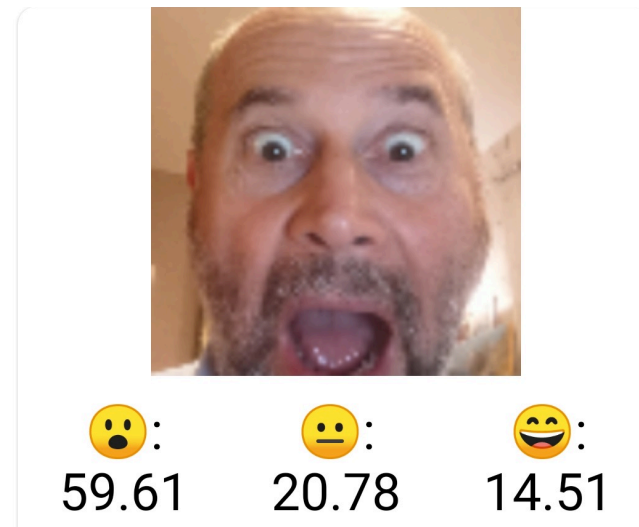
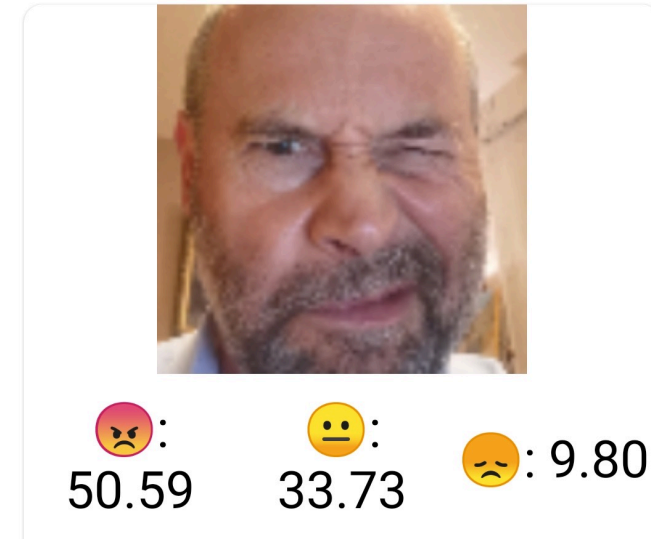
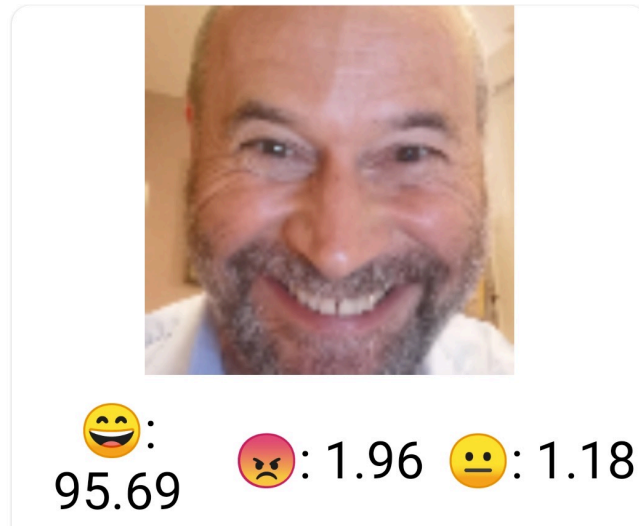
What is common
between all these
faces?



Find a subset of TF sites out of all 5260 TFs

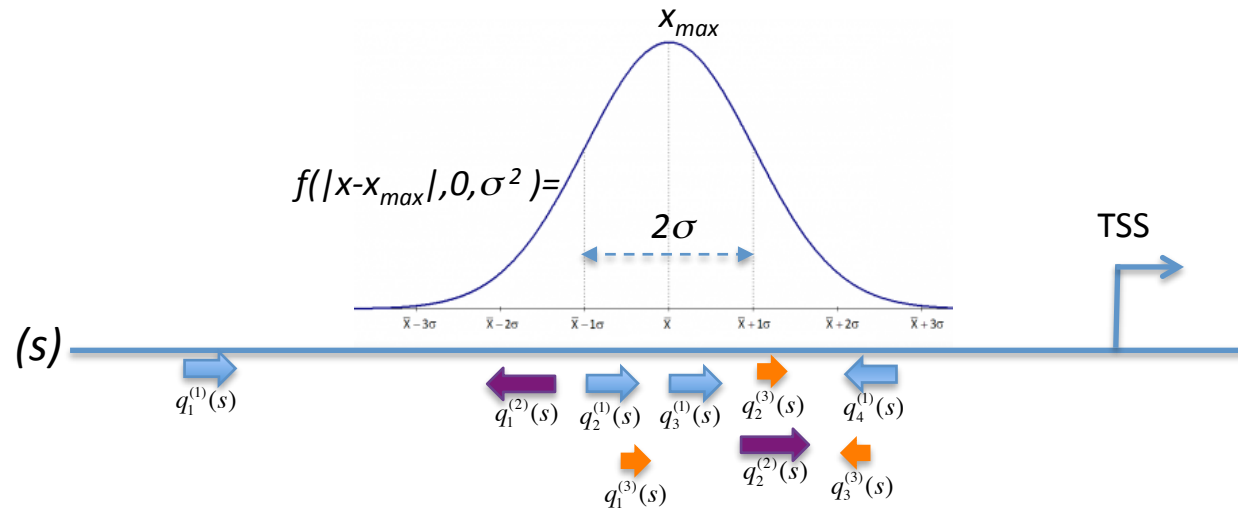


There are new AI methods available for recognition of human emotions:

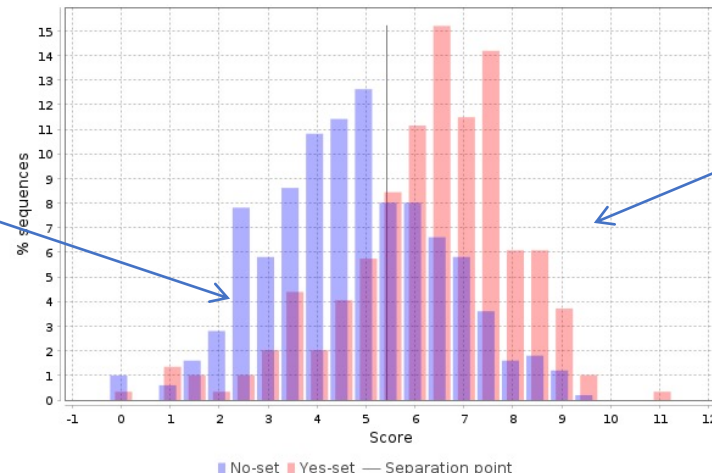


We are going to use such algorithms for recognizing promoter “emotions” – promoter “smile” or PD enhancer “grin”.

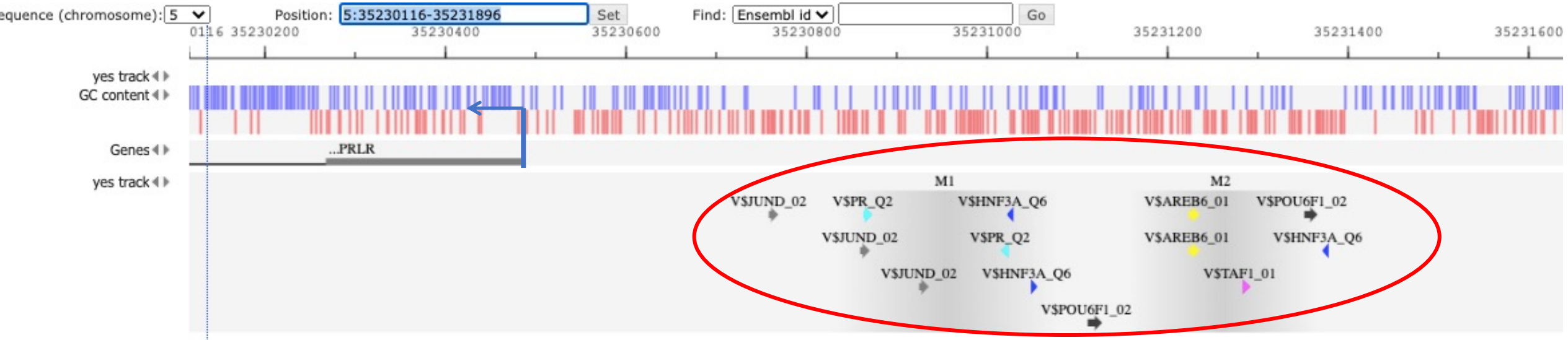
Composite model



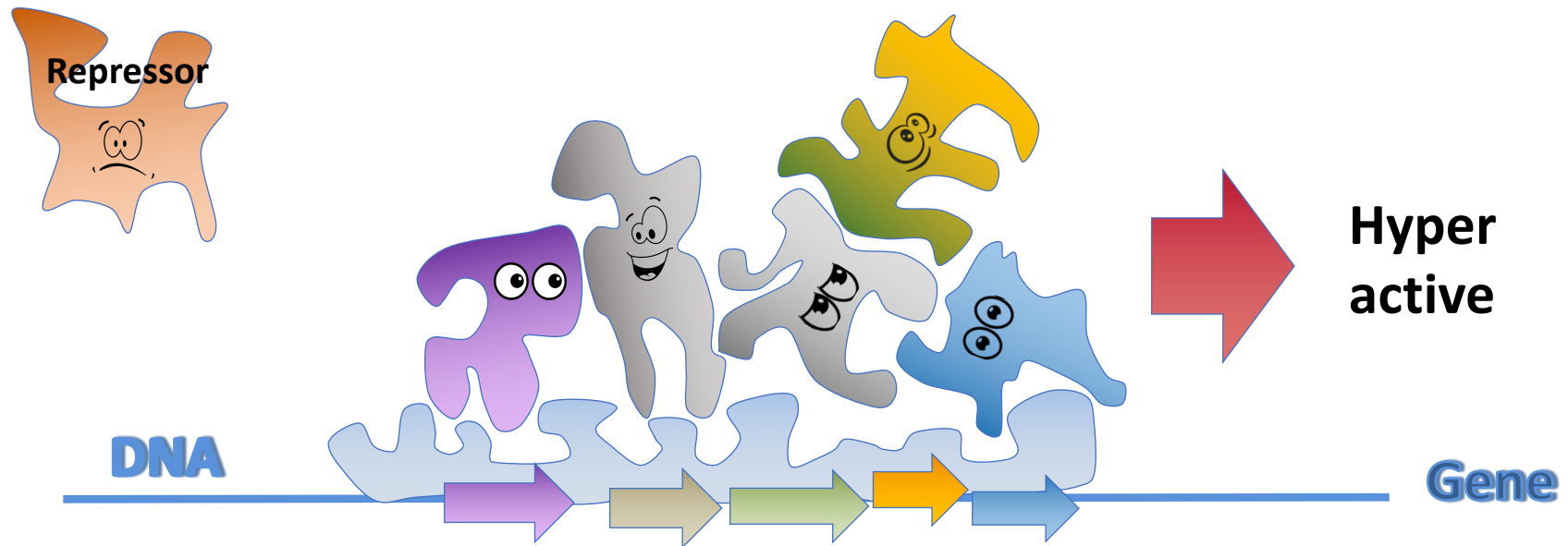
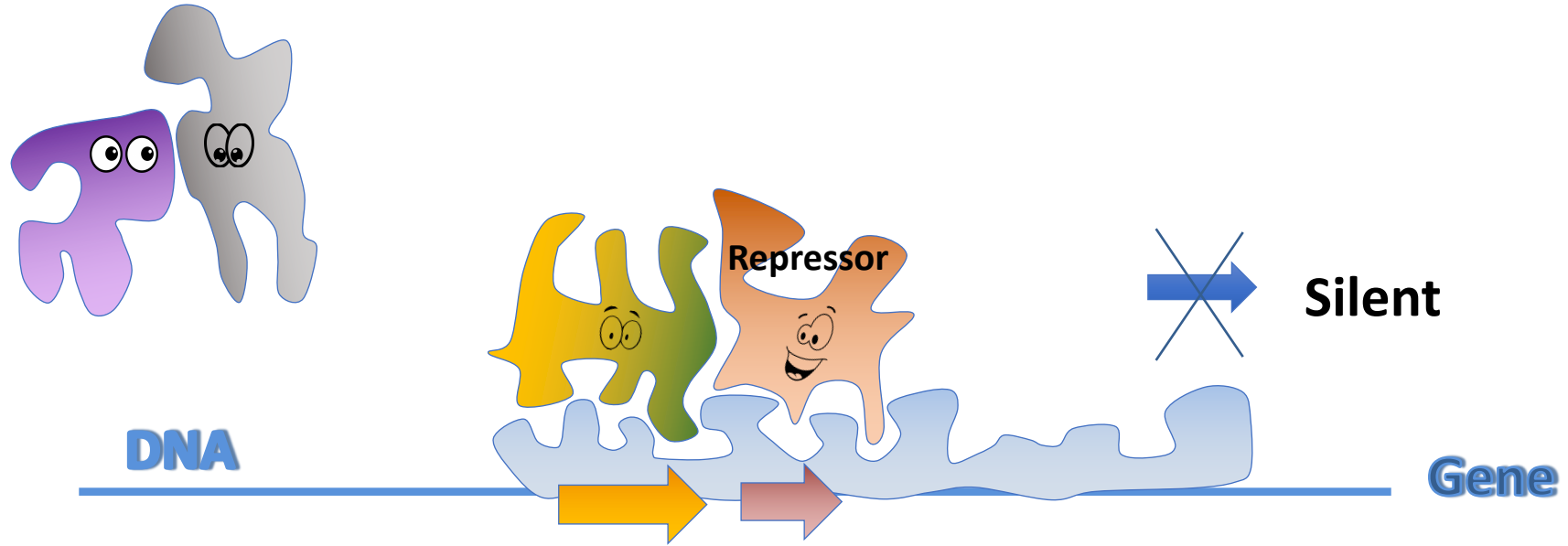
500
Non changed
genes



296 top
UP-regulated
genes
with TF
clusters

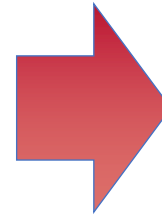
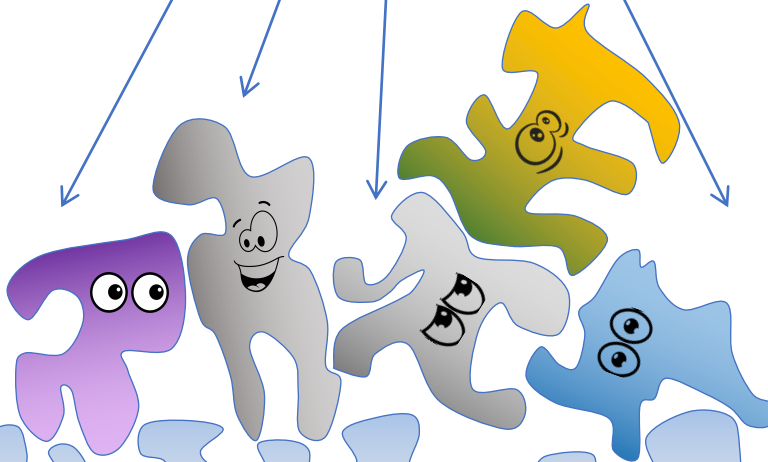


It's a fuzzy puzzle!





Master regulator ?



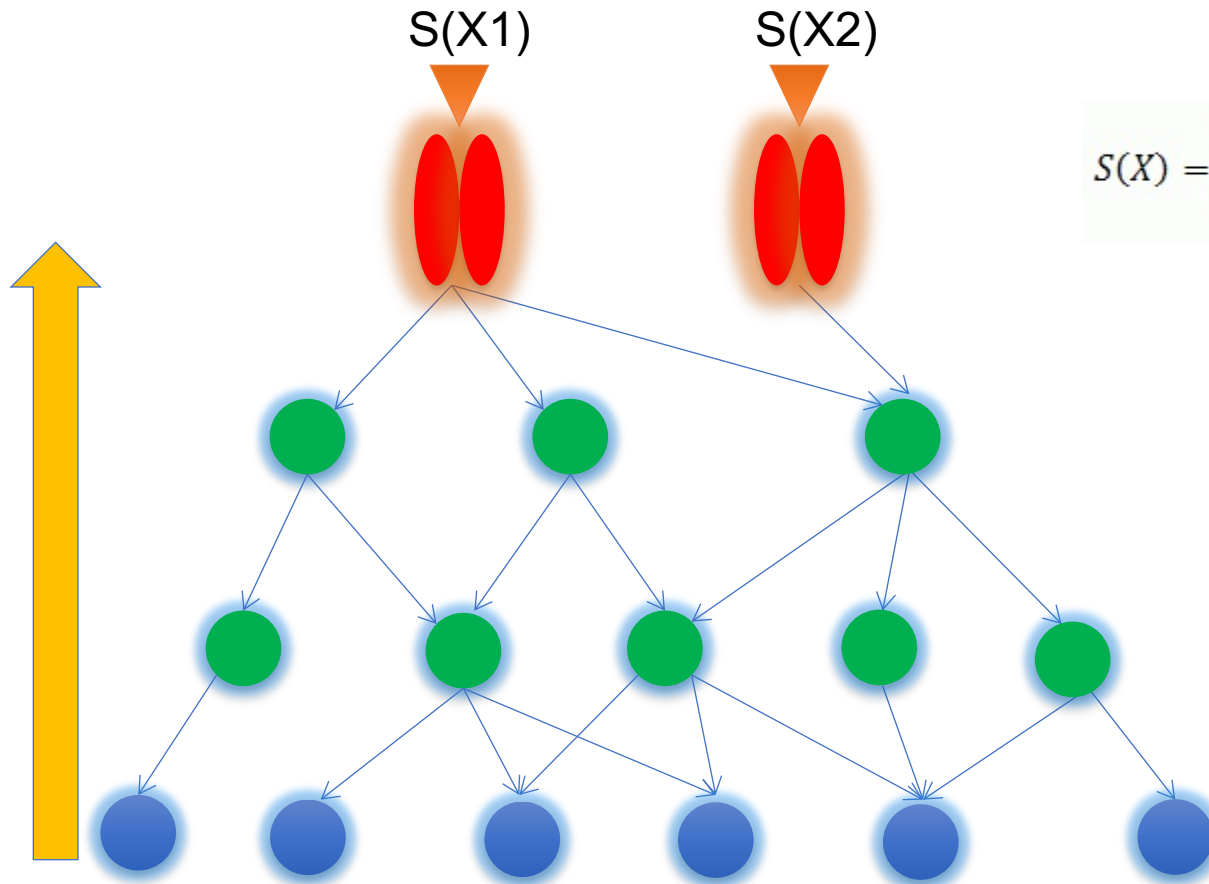
Hyper active

DNA

Gene



Search for master regulators



$$S(X) = \sum_{r=1}^R \frac{M(X,r)}{M_{max}(r)} \cdot \frac{1}{1 + pN(X,r)/N_{max}(r)}$$

Where:

R - Max radius (input parameter)

p - Penalty (input parameter)

N(X,r) - total number of molecules reachable from key molecule X within the radius r.

N_{max}(r) - maximal value of N(X,r) over all key molecules X found for this radius.

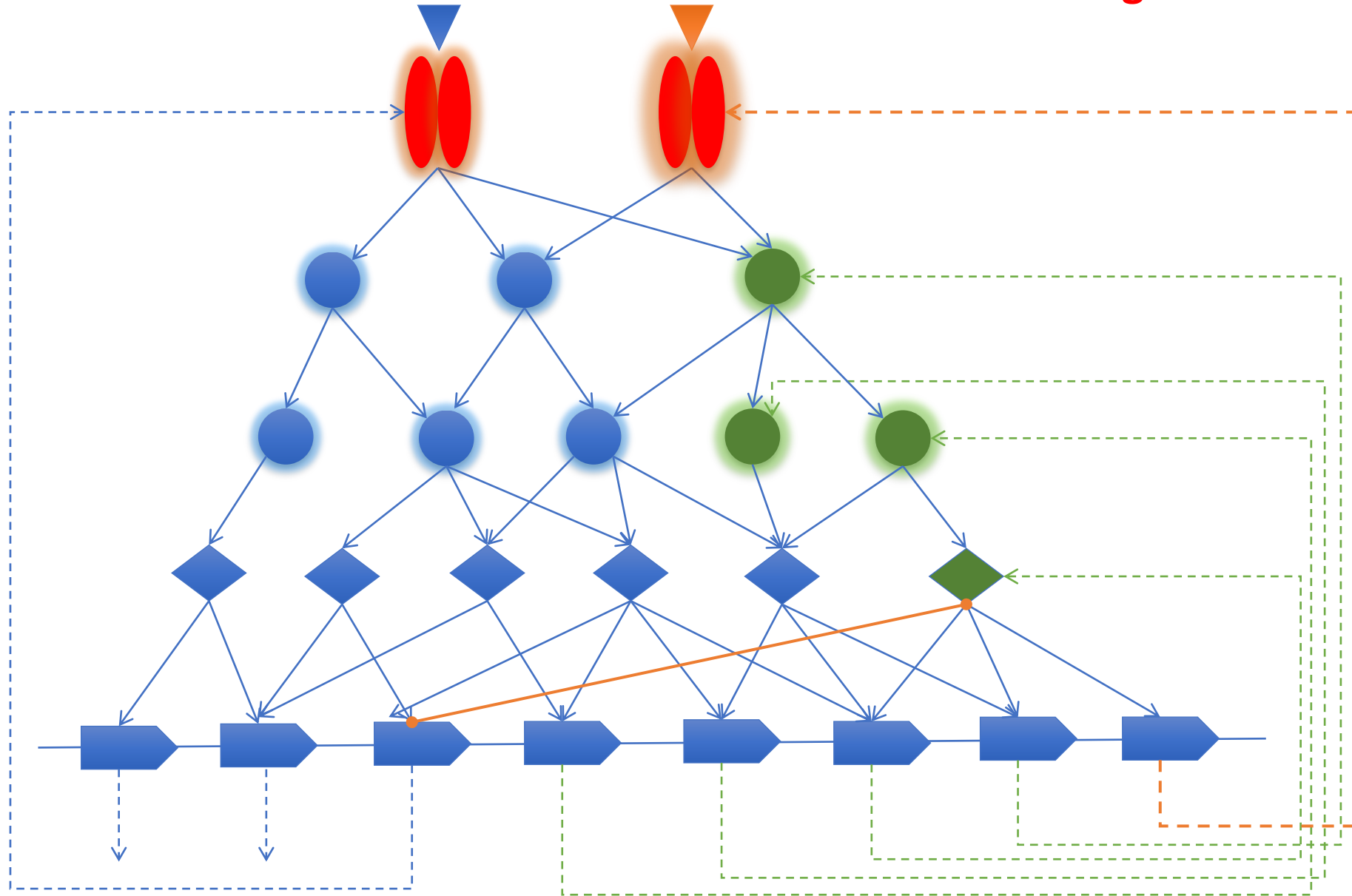
M(X,r) - sum of w(X) for all hits reachable from key molecule X within the radius r, where w(X) - weight of hit X.

M_{max}(r) - maximal value of M(X,r) over all key molecules X found for this radius.

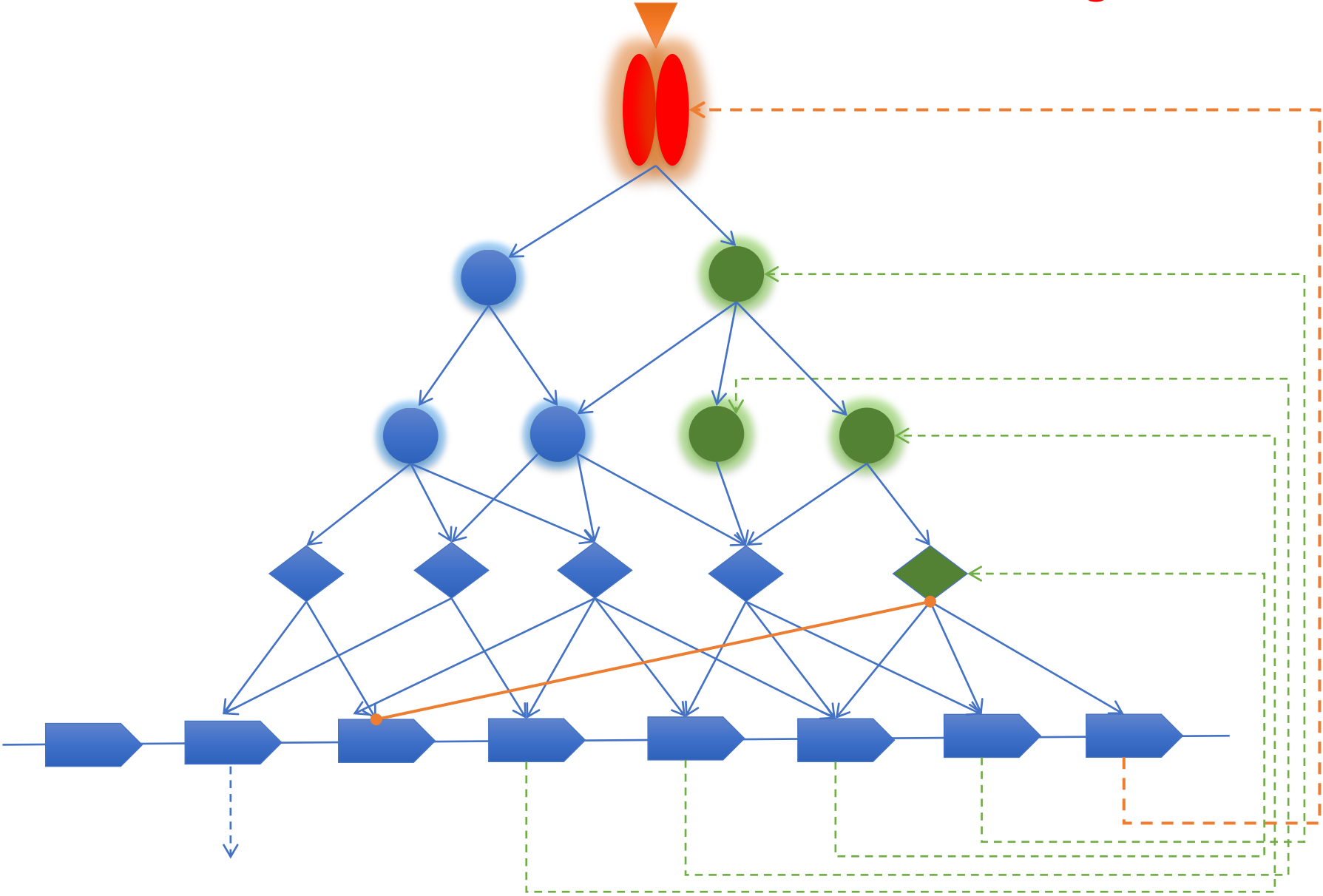
Kel, A., Voss, N., Jauregui, R., Kel-Margoulis, O. and Wingender, E.: Beyond microarrays: Find key transcription factors controlling signal transduction pathways BMC Bioinformatics 7(Suppl. 2), S13 (2006).

Master-regulator

New master-regulator

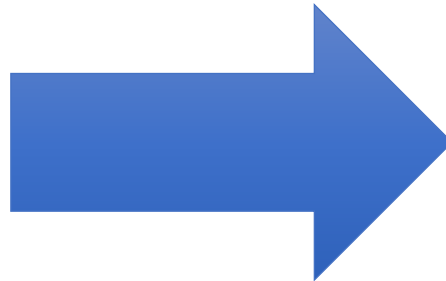
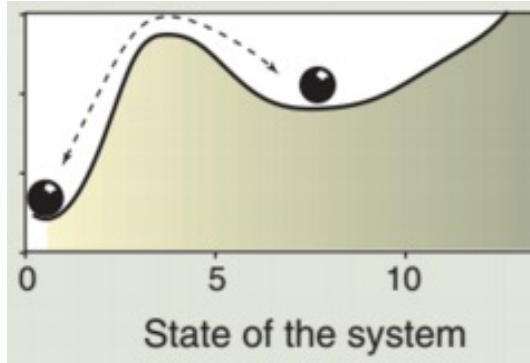
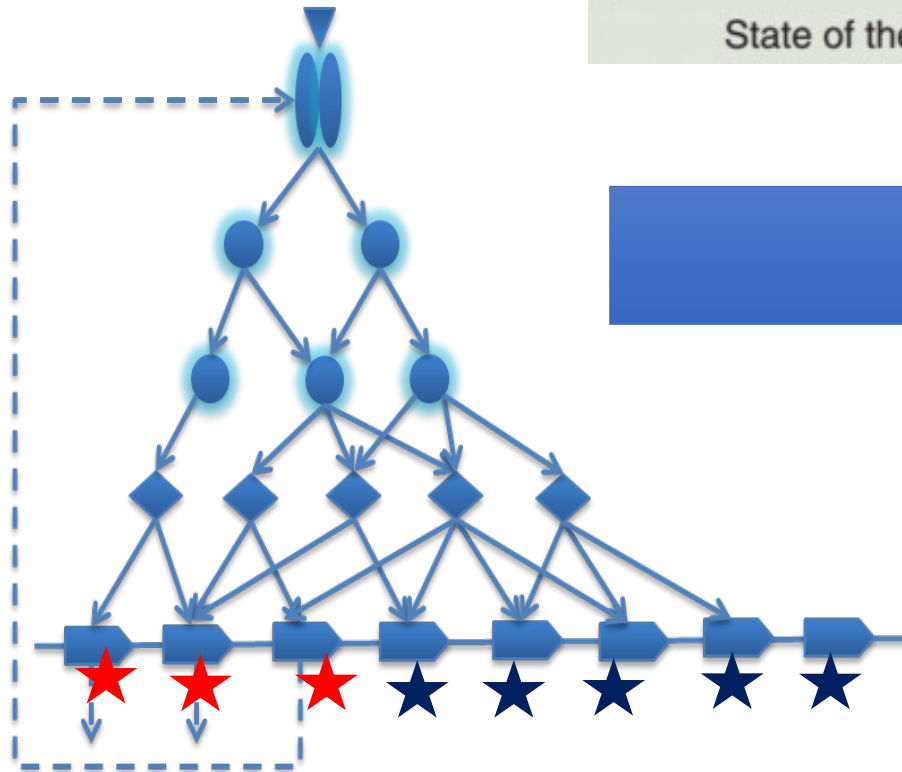


New master-regulator

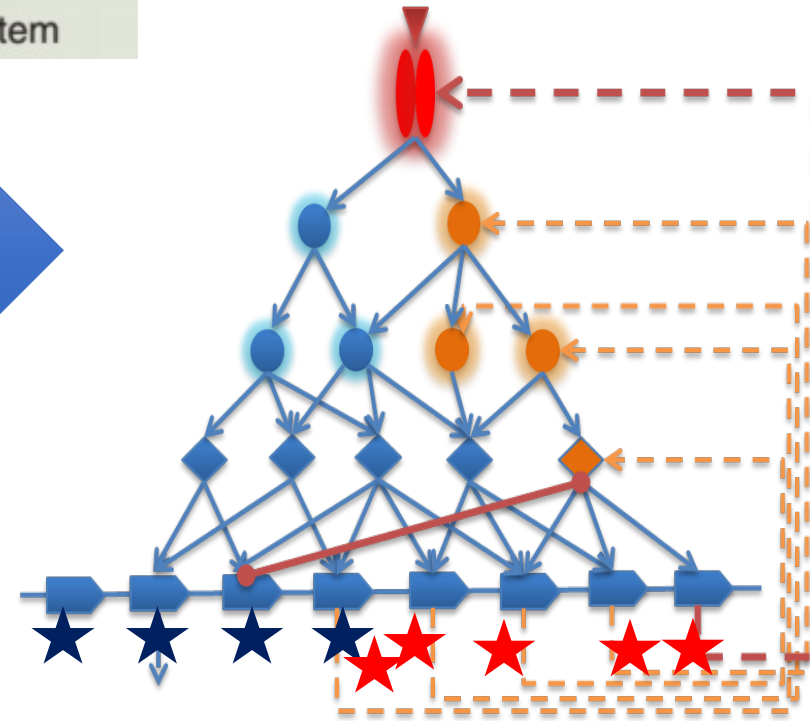


Pathway corruption

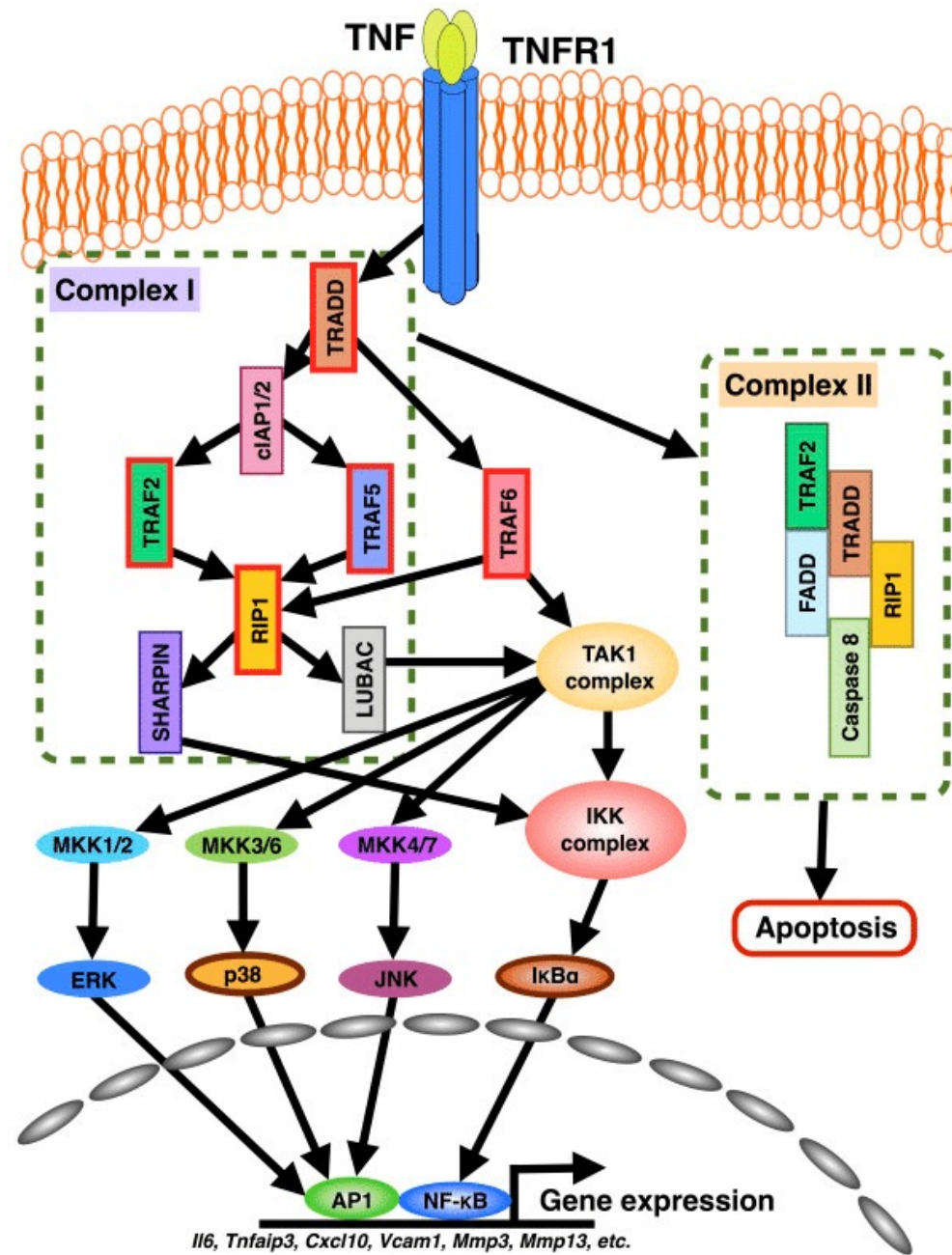
Healthy



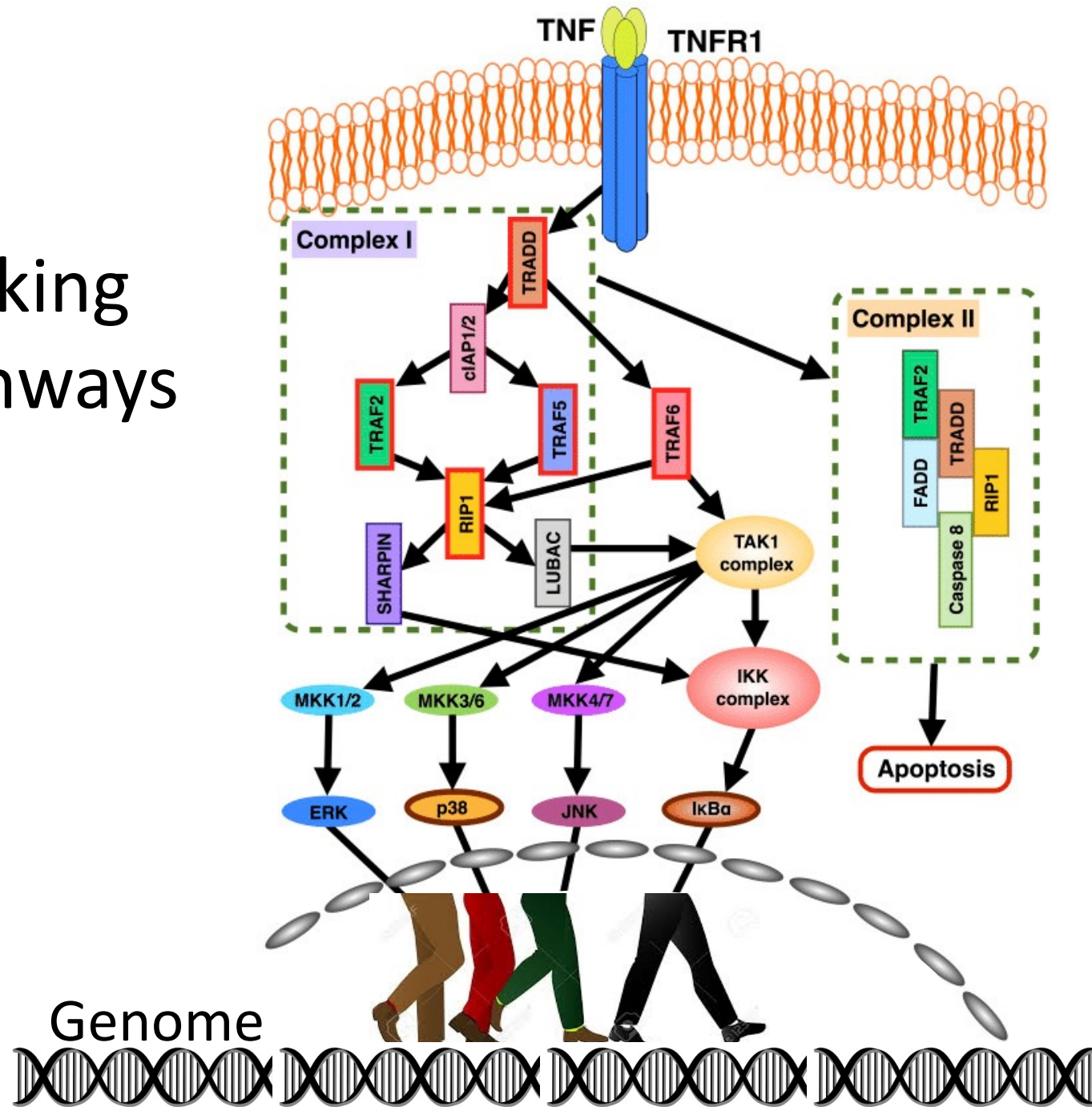
Disease



Walking pathways



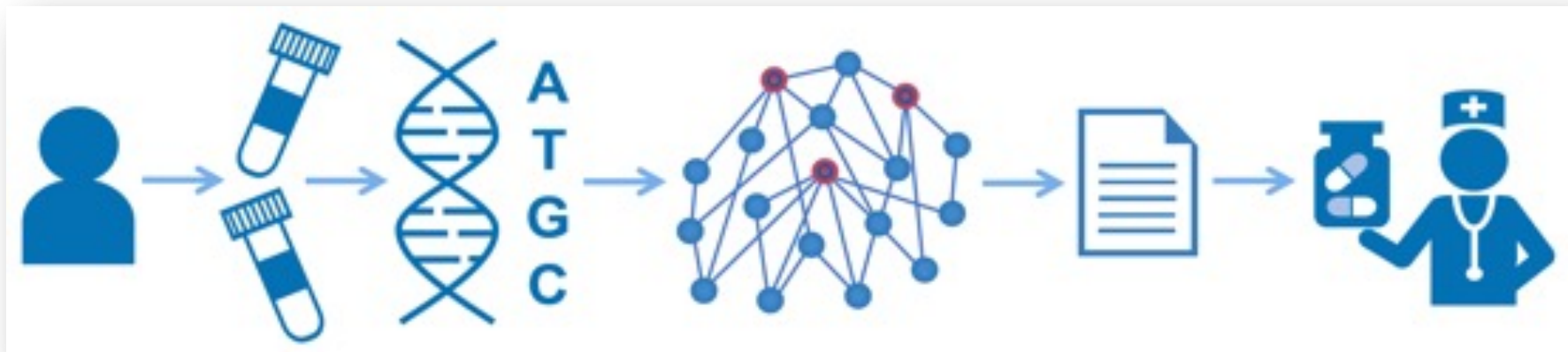
Walking pathways

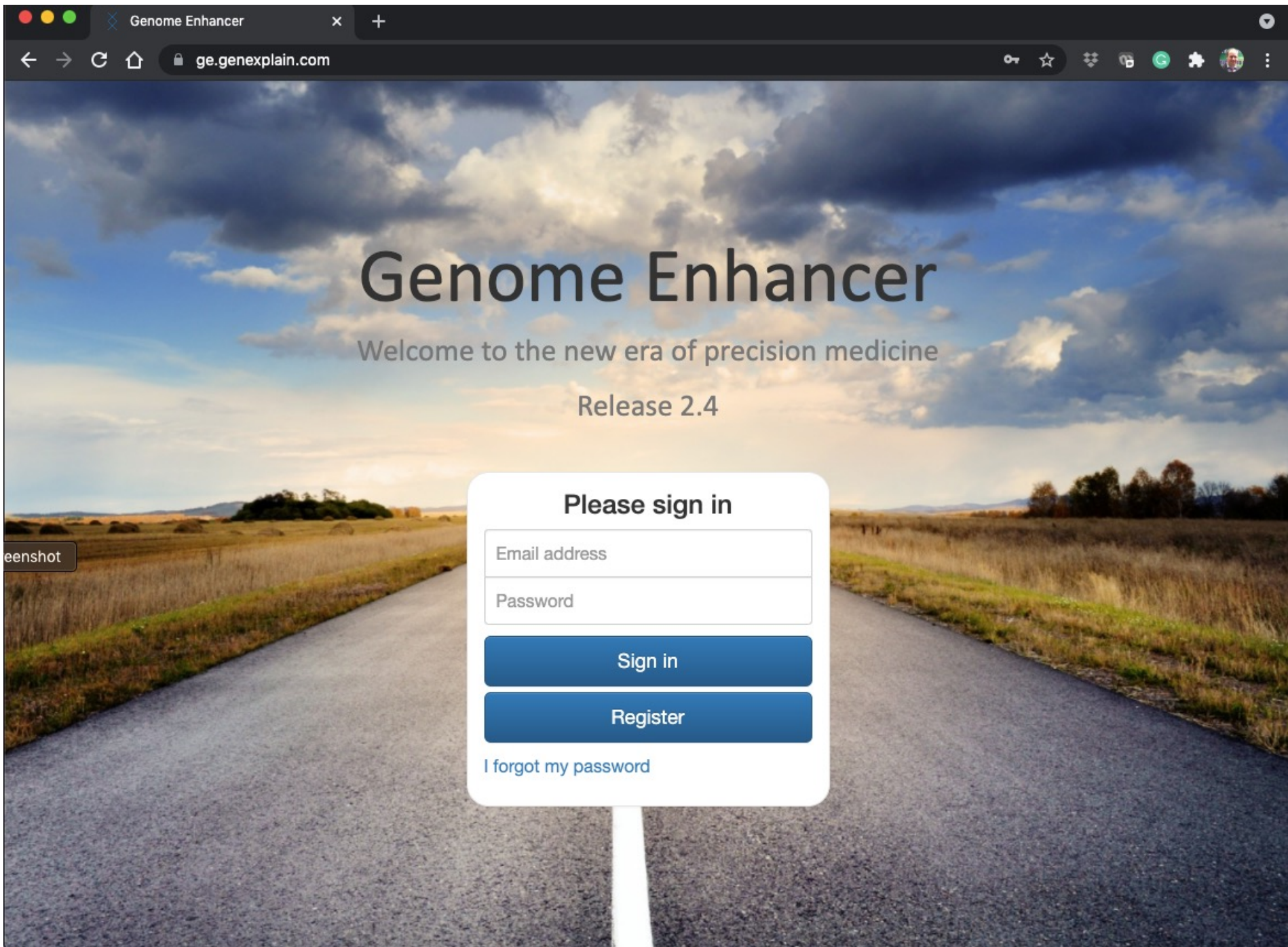


We would like AI to run the full data analysis completely automatic...



Genome Enhancer





Genome Enhancer

Welcome to the new era of precision medicine

Release 2.4

Please sign in

Sign in

Register

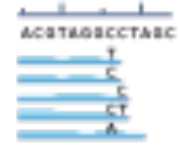
[I forgot my password](#)

Multi-omics data input

Transcriptomics



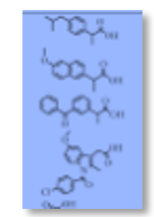
Genomics



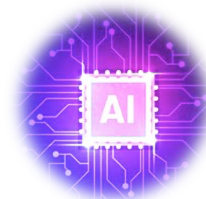
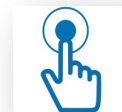
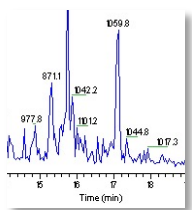
Epigenomics



Metabolomics

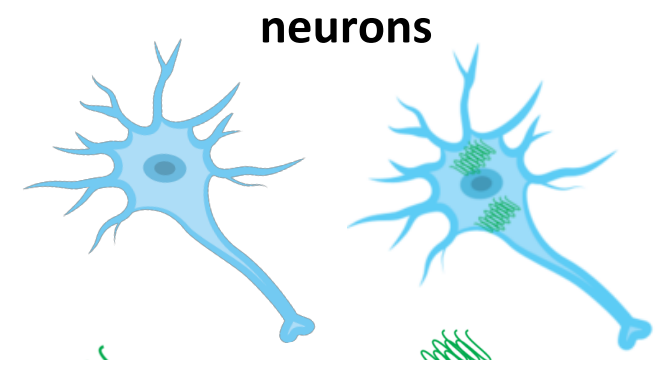


Proteomics



Status	Public on Feb 24, 2020
Title	α -Synuclein Translocates To The Nucleus To Activate Retinoic Acid- Dependent Gene Transcription
Organism	Homo sapiens
Experiment type	Expression profiling by high throughput sequencing
Summary	α -Synuclein (α -Syn) is a protein implicated in the pathogenesis of Parkinson's disease (PD). It is primarily cytosolic and reversibly interacts with cell membranes. α -Syn also occurs in the nucleus, however, the mechanisms involved in its nuclear localization are poorly understood. We analyzed alterations in gene expression following induced α -Syn expression in SH-SY5Y cells. Analysis for upstream regulators pointed at alterations in transcription activity of retinoic acid receptors (RAR)s and additional nuclear receptors. We show that α -Syn binds RA and translocates to the nucleus to selectively enhance gene transcription. Nuclear translocation of α -Syn is regulated by calreticulin, in a leptomycin-B independent mechanism. Importantly, nuclear translocation of α -Syn following RA treatment enhances its toxicity in cultured neurons and the expression levels of PD-associated genes, among which are two familial PD-associated genes, ATPase cation transporting 13A2 (ATP13A2) and PTEN-induced kinase 1 (PINK1). The results link a physiological role for α -Syn in the regulation of RA-mediated gene transcription and its toxicity in the synucleinopathies.
Overall design	SH-SY5Y cells induced to express α -Syn (with 1 μ g/ml doxycycline) and treated for 16 hours in DMEM supplemented with 0.1% serum and RA (5 μ M). Cells were collected and analyzed following 72 hours from induction of α -Syn expression. Control samples included an identical set up but without induced α -Syn expression.
Contributor(s)	Davidi D , Schechter M , Abd ElHadi S , Matatov A , Nathanson L , Sharon R
Citation missing	<i>Has this study been published? Please login to update or notify GEO.</i>
Submission date	Feb 24, 2020
Last update date	Feb 25, 2020
Contact name	Lubov Nathanson
E-mail(s)	lnathanson@nova.edu
Phone	954-262-2872
Organization name	Nova Southeastern University
Department	INIM
Street address	3321 College Ave.
City	Fort Lauderdale
State/province	Florida
ZIP/Postal code	33314

Alpha-Syn



2. Data

For this study the following experimental data was used:

Table 1. Experimental datasets used in the study

File name	Data type
GSE145804_DESeq2_final.csv/GSE145804_DESeq2_final	Transcriptomics

noRA_Dox	
raw.S16	GSE145804_DESeq2_final
raw.S17	GSE145804_DESeq2_final
raw.S18	GSE145804_DESeq2_final

noRA_noDox	
raw.S13	GSE145804_DESeq2_final
raw.S14	GSE145804_DESeq2_final
raw.S15	GSE145804_DESeq2_final

detected 8084 upregulated genes (LogFC>0) out of which 578 genes were found as significantly upregulated (p-value<0.1) and 8862 downregulated genes (LogFC<0) out of which 726 genes were significantly downregulated (p-value<0.1). See tables below for the top significantly up- and downregulated genes. Below we call **target genes** the full list of up- and downregulated genes revealed in our analysis (see tables in [Supplementary section](#)).

Table 2. Top ten significant **up-regulated** genes in *noRA_Dox* vs. *noRA_noDox*.

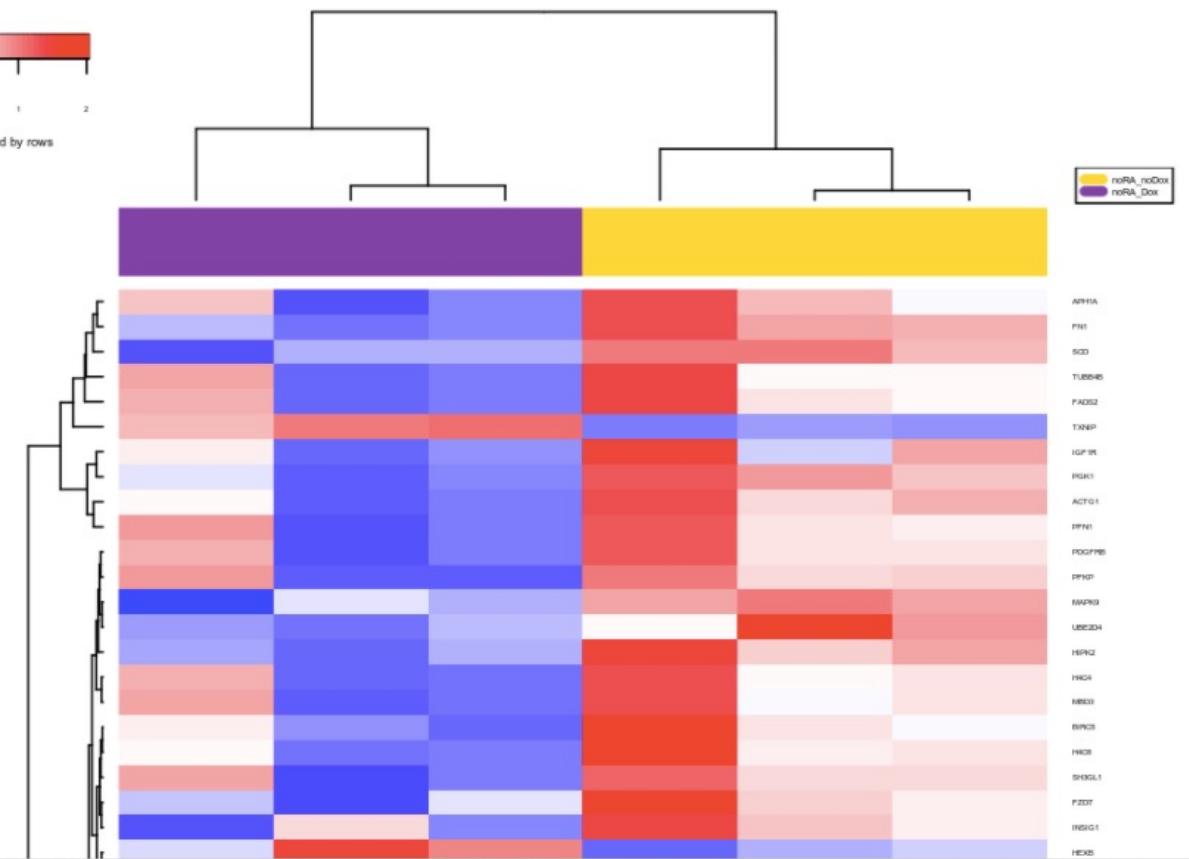
[See full table](#) →

ID	Gene symbol	Gene description	logFC	logCPM	PValue	FDR
ENSG00000145335	SNCA	synuclein alpha	4.38	10.36	2.71E-88	4.59E-84
ENSG00000118785	SPP1	secreted phosphoprotein 1	2.98	-0.28	6.23E-7	2.11E-3
ENSG00000169282	KCNAB1	potassium voltage-gated channel subfamily A member regulatory beta subunit 1	2.91	0.49	1.32E-7	5.61E-4
ENSG00000162692	VCAM1	vascular cell adhesion molecule 1	2.67	-4.05E-2	1.1E-5	2.34E-2
ENSG00000237280	AL136982.3	novel transcript	2.15	0.2	1.42E-4	0.16
ENSG00000214892	USP8P1	ubiquitin specific peptidase 8 pseudogene 1	1.81	0.19	4.4E-3	0.83
ENSG00000224837	GCSHP5	glycine cleavage system protein H pseudogene 5	1.72	-0.1	1.84E-3	0.6
ENSG00000243300	null	null	1.65	0.44	2.8E-3	0.68
ENSG00000229474	PATL2	PAT1 homolog 2	1.53	-0.14	7.97E-3	0.98
ENSG00000236813	BTF3P8	basic transcription factor 3 pseudogene 8	1.53	-0.48	9.95E-3	0.98

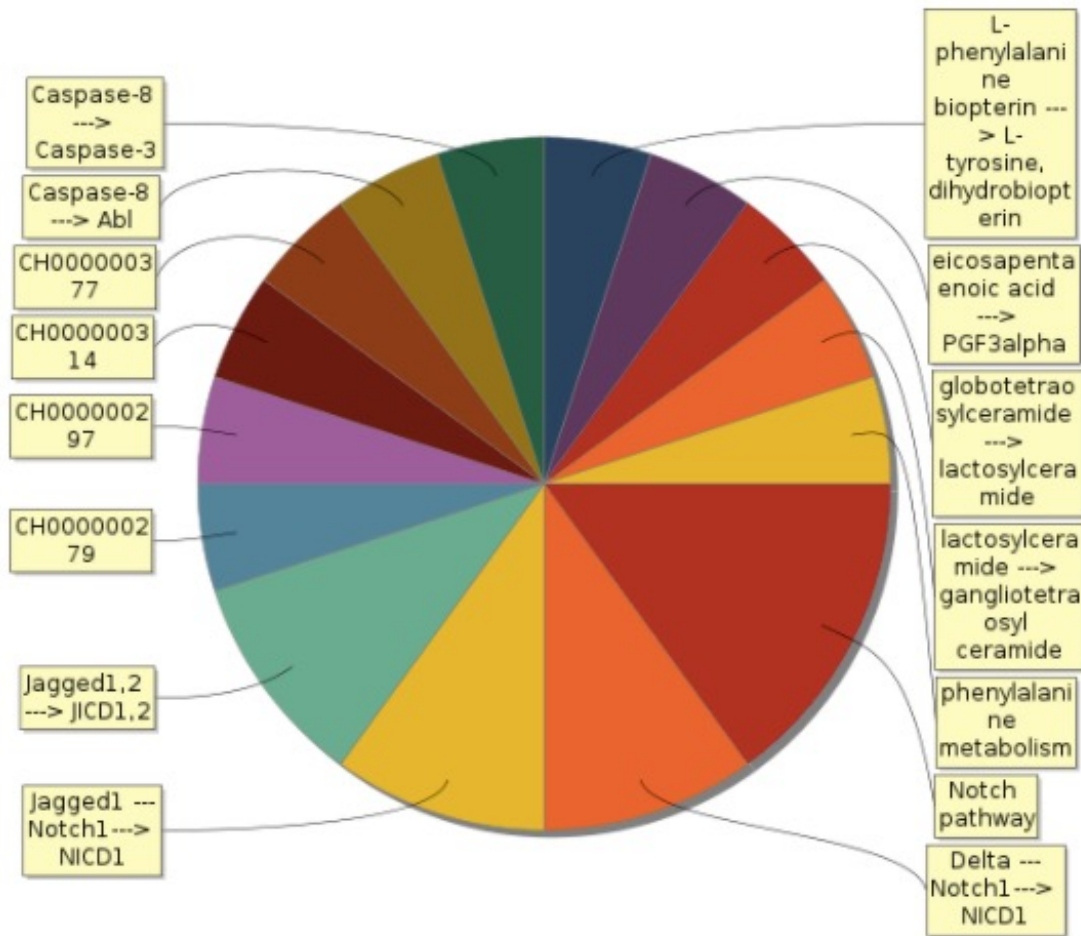
Table 3. Top ten significant **down-regulated** genes in *noRA_Dox* vs. *noRA_noDox*.

[See full table](#) →

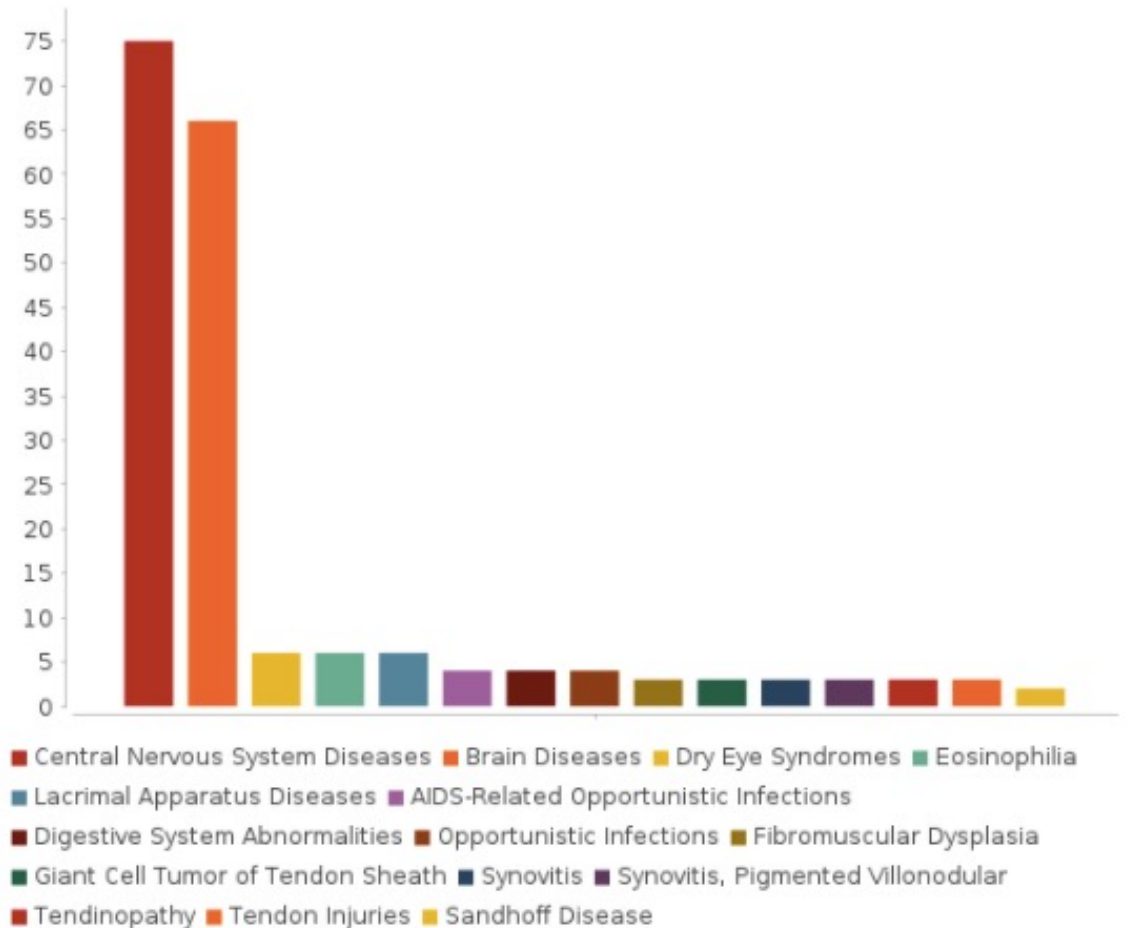
ID	Gene symbol	Gene description	logFC	logCPM	PValue	FDR
ENSG00000186081	KRT5	keratin 5	-10.36	2.57	9.35E-7	2.64E-3
ENSG00000257594	GALNT4	polypeptide N-acetylgalactosaminyltransferase 4	-2.37	1.39E-2	4.31E-5	6.08E-2
ENSG00000167244	IGF2	insulin like growth factor 2	-2.13	-0.49	1.23E-3	0.47
ENSG00000255115	AP002812.4	family with sequence similarity 162, member A (FAM162A) pseudogene	-1.97	-0.38	2.22E-3	0.61
ENSG00000134855	SLC27A2	solute carrier family 37 member	-1.8	-0.42	9.67E-3	0.98



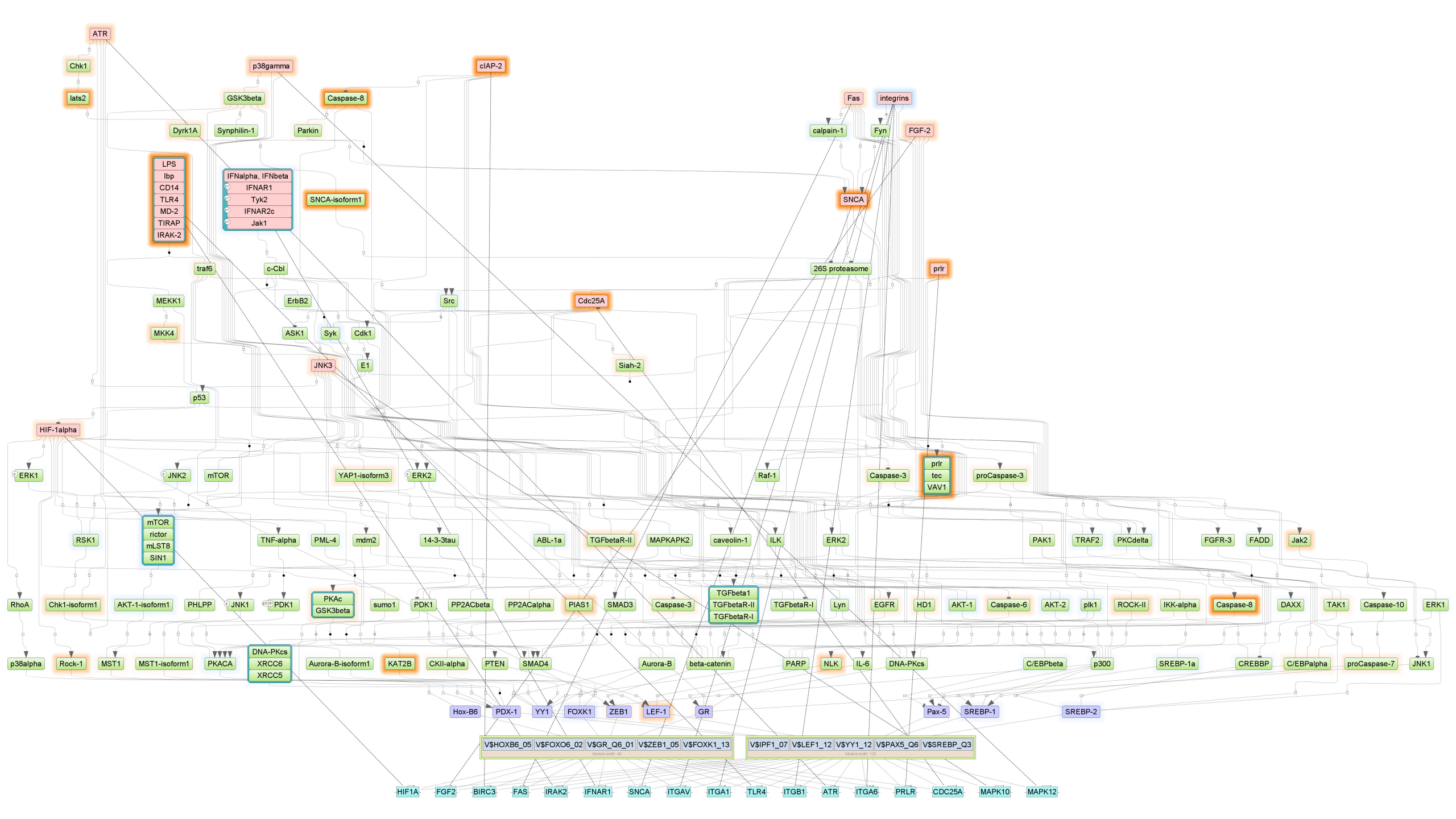
TRANSPATH® Pathways (2020.2)

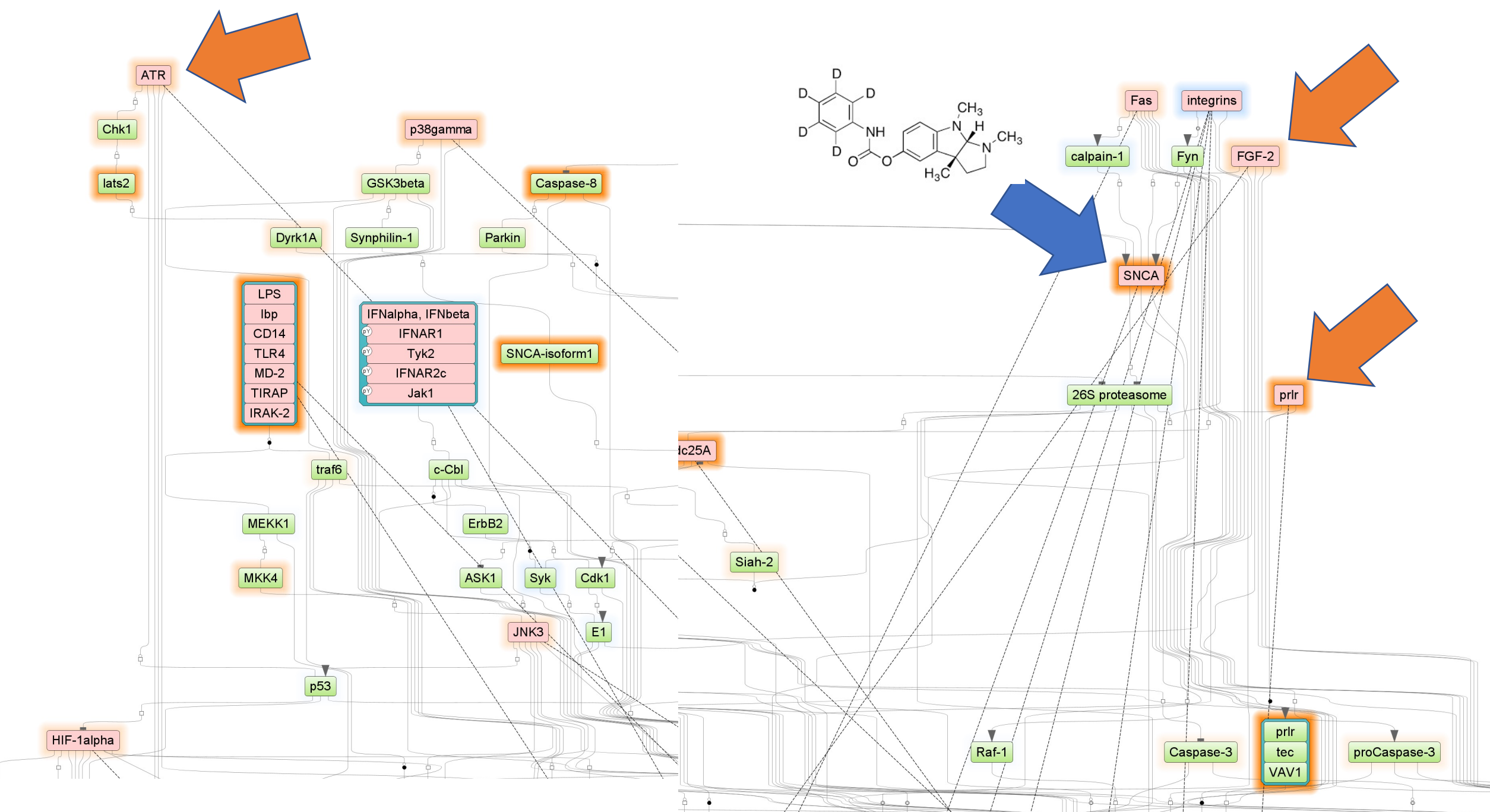


HumanPSD(TM) disease (2020.2)









Drugs approved in clinical trials



Table 10. FDA approved drugs or drugs used in clinical trials for the studied pathology (most promising treatment candidates selected for the identified drug targets on the basis of literature curation in *HumanPSD™* database)

[See full table](#) →

Name	Target names	Drug score	Disease activity score	Disease trial phase
Sirolimus	IKBKB, MAPK10, RPS6KA3, ROCK2, TGM2, MAPK12, DYRK1A... (more)	86	3	Phase 2: Parkinson Disease, Acute Disease, Acute Lung Injury, Adenocarcinoma, Adenocarcinoma of Lung... (more)
Caffeine	ATR, FGF2	64	6	Phase 3: Parkinson Disease, Alzheimer Disease, Apnea, Asymptomatic Diseases, Atrial Fibrillation, Br... (more)
Acetylsalicylic acid	IKBKB, FAS, FGF2	53	4	Phase 4: Parkinson Disease, Acute Coronary Syndrome, Aneurysm, Aneurysm, Dissecting, Angina Pectoris... (more)
Cholecalciferol	FAS	43	7	Phase 4: Parkinson Disease, Affect, Amenorrhea, Anemia, Anemia, Aplastic, Anemia, Sickle Cell, Arthr... (more)
Melatonin	PRLR	37	9	Phase 4: Parkinson Disease, Arteriosclerosis, Attention Deficit Disorder with Hyperactivity, Autism ... (more)

The ***Disease trial phase*** column reflects the maximum clinical trials phase in which the drug was studied for the analyzed pathology.

Repurposing drugs



Table 11. Repurposed drugs used in clinical trials for other pathologies (prospective drugs against the identified drug targets on the basis of literature curation in *HumanPSD™* database)

[See full table](#) →

Name	Target names	Drug score	Maximum trial phase
1-(5-Tert-Butyl-2-P-Tolyl-2h-Pyrazol-3-Yl)-3-[4-(2-Morpholin-4-Yl-Ethoxy)-Naphthalen-1-Yl]-Urea	IKBKB, MAPK10, RPS6KA3, TEC, ROCK2, PRKCQ, TTK... (more)	94	N/A
Sorafenib	IKBKB, MAPK10, RPS6KA3, TEC, ROCK2, PRKCQ, TTK... (more)	94	Phase 4: Carcinoma, Carcinoma, Hepatocellular, Carcinoma, Renal Cell, Hand-Foot Syndrome, Liver Neop... (more)
seliciclib	IKBKB, MAPK10, RPS6KA3, TEC, ROCK2, PRKCQ, TTK... (more)	94	Phase 2: ACTH-Secreting Pituitary Adenoma, Adenoma, Carcinoma, Non-Small-Cell Lung, Cystic Fibrosis,... (more)
ruboxistaurin	IKBKB, MAPK10, RPS6KA3, TEC, ROCK2, PRKCQ, TTK... (more)	94	Phase 3: Diabetes Mellitus, Diabetes Mellitus, Type 1, Diabetes Mellitus, Type 2, Diabetic Neuropath... (more)
Tofacitinib	MAPK10, RPS6KA3, TEC, ROCK2, RIPK2, MERTK, LATS2... (more)	93	Phase 4: Alopecia, Alopecia Areata, Aortic Arch Syndromes, Arteritis, Arthritis, Arthritis, Psoriati... (more)

The **Maximum trial phase** column reflects the maximum clinical trials phase in which the drug was studied for any pathology.



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ANTI-PARKINSON'S ACTIVITY OF SORAFENIB IN 6-OHDA INDUCED RAT MODEL

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Keywords:

6-OHDA, Levodopa,
Neuroprotective, Sorafenib

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ABSTRACT: Objective: Studies have shown that sorafenib an anti-cancer agent has a neuroprotective effect. This study evaluated the neuroprotective activity of sorafenib in 6-OHDA induced rat model of Parkinson's disease. **Methods:** 6-OHDA was injected into the forebrain bundle through the stereotaxic apparatus to induce fast and severe degeneration in dopaminergic neurons of substantianigra. The animals were divided into four groups. Group I- vehicle control, Group II- 6-OHDA induced, Group III- 6- OHDA + Levodopa (6 mg/kg), Group IV- 6-OHDA + sorafenib (10 mg/kg s.c). Treatment was given for 21 days after induction of 6-OHDA. The animals were subjected to behavioral parameters such as apomorphine-induced rotations, grip strength, catatonia and biochemical parameters such as total protein estimation, reduced glutathione, lipid peroxidase, calcium concentration in the brain. **Results:** Sorafenib significantly decreased the apomorphine-induced rotations as well as catatonia and significantly increased ($p < 0.001$) the grip strength when compared to 6-OHDA. In biochemical estimation total protein and glutathione is increased ($p < 0.001$). Both lipid peroxidase and calcium level have been decreased significantly ($p < 0.001$) when compared to 6. OHDA. **Conclusion:** In the present study, antiparkinson's effect of an LRRK2 inhibitor, sorafenib was evaluated in the 6-OHDA lesioned rat model. Behavioral and biochemical parameters were carried out. The parameters revealed that the LRRK2 inhibitor, sorafenib has shown significant antiparkinson's activity. The estimated parameters altered the normal behavior of the animal and the drug treatment protected the diseased brain of rat.

INTRODUCTION: Parkinson's disease (PD) is a progressive neurodegenerative disorder caused by PD is the second most common neurodegenerative disease after Alzheimer's disease affecting

PASS

The acronym PASS stands for Prediction of Activity Spectra for Substances. PASS performs an instant prediction and computational evaluation of biological activity spectra for organic chemical compounds.

PASS results can be further interpreted via the PharmaExpert tool and combined with the structure-activity relationship models built in GUSAR.

C:\Program Files (x86)\PASS\PASS 2011 Professional\Samples\Prediction Results\Drugs_Example (PASS11).SDF

5x5 | 4x4 | 3x3 | 2x2 | Molecular Structure | MNA |

64 65 66

67 68 69

70 71 72

68/128

No Selected Activity

Activity Map General Effects Mechanisms Toxicity Antitargets Metabolism

63 of 4366 Possible Activities at Pa > 0.500

- 0.818 0.010 Tachycardiac
- 0.776 0.004 Skeletal muscle relaxant
- 0.787 0.027 Polyporopepsin inhibitor
- 0.769 0.021 Hepatitis
- 0.733 0.011 Hypothermic
- 0.758 0.036 Acrocyliindropepsin inhibitor
- 0.758 0.036 Chymosin inhibitor
- 0.758 0.036 Saccharopepsin inhibitor
- 0.776 0.057 Anticemetic
- 0.715 0.004 Muscle relaxant
- 0.731 0.025 5-Hydroxytryptamine release stimulant
- 0.684 0.034 Antischismic, cerebral
- 0.660 0.019 Coma
- 0.674 0.036 Hypercholesterolemic
- 0.657 0.026 Consciousness alteration
- 0.665 0.038 CyP25B inhibitor
- 0.685 0.087 Antacid
- 0.648 0.057 Nicotinic alpha6beta3beta4alpha5 receptor antagonist
- 0.605 0.017 Antineoplastic (non-Hodgkin's lymphoma)

32 Substructure Descriptors: 0 new.

63 of 4366 Possible Activities

- 12 of 497 Possible Pharmacological Effects
- 22 of 3378 Possible Mechanisms of Action
- 24 of 274 Possible Toxic and Adverse Effects
- 1 of 116 Possible Antitargets
- 5 of 206 Possible Metabolism-Related Actions
- 0 of 31 Possible Gene Expression Regulation
- 0 of 49 Possible Transporters-Related Actions

C:\Program Files (x86)\PASS\PASS 2011 Professional\Samples\Prediction Results\Drugs_Example (PASS11).SDF

5x5 | 4x4 | 3x3 | 2x2 | Molecular Structure | MNA |

67 68 69

70

68/128

Fibrinolytic

Activity Map General Effects Mechanisms Toxicity Antitargets Metabolism

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- 5 of 206 Possible Metabolism-Related Actions
- 0 of 31 Possible Gene Expression Regulation

> <NAME> (68)
Metaxalone

0.322 0.199 Fibrinolytic

Select Activity Types to be Predicted

Find: Select: Sort: IEP ascending

Predictable Activity Type	Group	Number	IEP, %
Hydroxylsine kinase inhibitor	M	9	0.336
Endothelin A receptor antagonist	M	1715	0.336
Histone deacetylase 1 inhibitor	M	326	0.338
Glutamate formimidoyltransferase inhibitor	M	11	0.339
Melanocortin MC-4 agonist	M	15	0.343
Hydroxyquinol 1,2-dioxygenase inhibitor	M	9	0.343
Glucosamine N-acetyltransferase inhibitor	M	6	0.344
Retinoid X receptor agonist	M	112	0.344
[glutamate-ammonia-lyase] adenylyltransferase inhibitor	M	13	0.346
Rotamase (FKBP) inhibitor	M	6	0.346

Unused Activity Type	Group	Number	IEP, %
Aggression	T	46	25.009
Keratitis	T	7	25.056
Dysphagia	T	15	25.352
Skin lesion	T	7	25.393
Bundle branch block	T	14	25.404
Agitation	T	60	25.414
Movement disorder	T	9	25.529
Coronary artery spasm	T	8	25.533
Pruritus	T	121	25.600
Allergic reaction	T	226	25.756

Include ... Load ... Save ... Ok Cancel

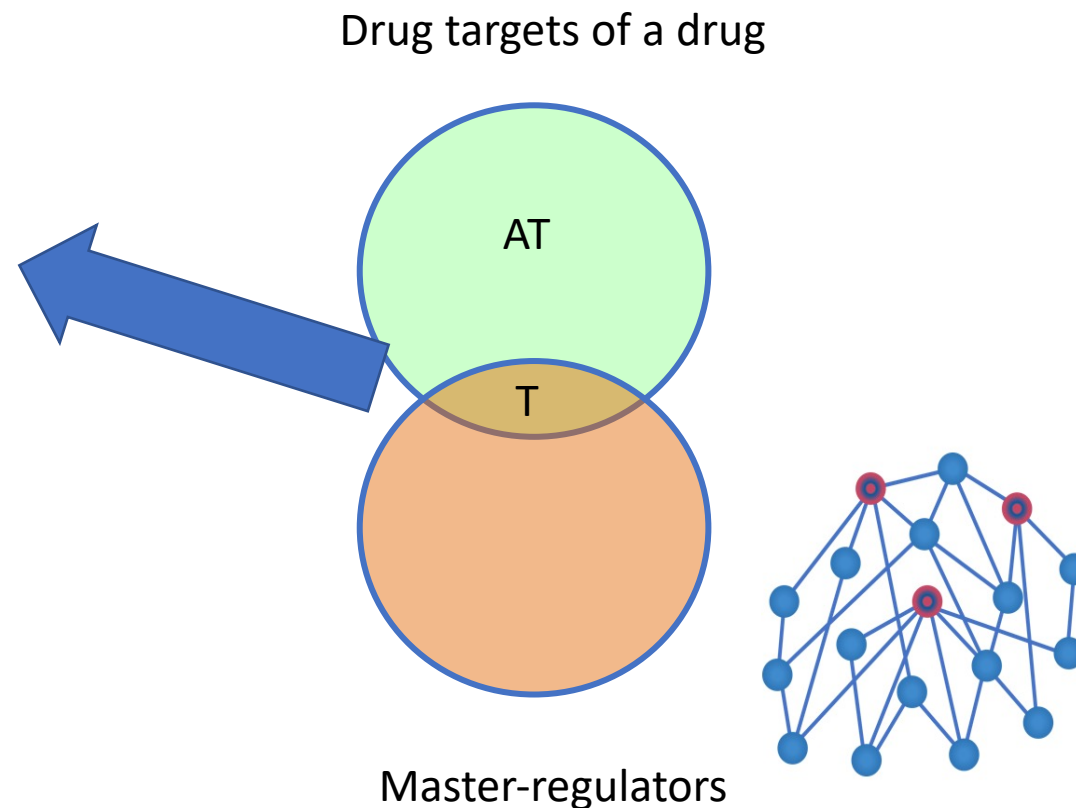
Target activity score



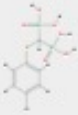






$$T\text{-score}(s) = \frac{|T|}{|T| + w(|AT| - |T|)} \sum_{m \in M(s)} \left(pa(m) \sum_{g \in G(m)} IAP(g) optWeight(g) \right),$$

$M(s)$ is the set of activity-mechanisms for the given structure s

$G(m)$ is the set of targets (converted to genes)

$IAP(g)$ is the invariant accuracy of prediction



Name	Structure	Target names	Target activity score	Disease activity score	Disease activity rank	Drug score
<u>Tacrolimus</u>		PPM1M, PPM1B, IL7, PPM1G, GDNF, BDNF, PPM1D, NEK6, FLT1, IL1A, IL2RB, IL10	1.0298	0.996	1	96
<u>Lipoic Acid</u>		PTPRO, PTPN1, PTPRC, PTPN22, PTPRE, UBASH3B, PTPN14, DUSP4, PPM1M, PPM1B, CDC25A, DUSP2, DUSP5, ... (more)	1.44298	0.942	59	95
<u>Tiludronate</u>		IL7, PTPRO, PTPN1, IL1A, PTPRC, PTPN22, IL10, PTPRE, UBASH3B, PTPN14, DUSP4, PPM1M, PPM1B, CDC25A... (more)	1.41195	0.937	63	95
<u>Fluocinolone Acetonide</u>		IL7, TLR3, TLR4, HIF1A, IL1A, IL10, RELA	0.34081	0.993	2	93
<u>Fluocinonide</u>		IL7, TLR3, TLR4, HIF1A, IL1A, IL10, RELA	0.33469	0.992	3	93
<u>Dexamethasone</u>		IL7, TLR3, TLR4, HIF1A, IL1A, IL10, RELA	0.31757	0.987	9	92
<u>Betamethasone</u>		IL7, TLR3, TLR4, HIF1A, IL1A, IL10, RELA	0.31757	0.987	9	92
<u>Diflorasone</u>		IL7, TLR3, TLR4, HIF1A, IL1A, IL10, RELA	0.30191	0.992	3	92
<u>Flumethasone Pivalate</u>		IL7, TLR3, TLR4, HIF1A, IL1A, IL10, RELA	0.3033	0.988	8	92
		PTPRO, PTPN1				

Alpha-L
Accumu
Parkins

Dexamethasone protects against dopaminergic neurons damage in a mouse model of Parkinson's disease

Shengyan Tai ¹
Chunlin Zhang

Iwona Kurkowska-Jastrzebska ¹, Tomasz Litwin, Ilona Joniec, Agnieszka Ciesielska, Adam Przybyłkowski, Andrzej Członkowski, Anna Członkowska

Affiliations +
PMID: 326700

Affiliations + expand

Free PMC artic

PMID: 15313429 DOI: 10.1016/j.intimp.2004.05.006

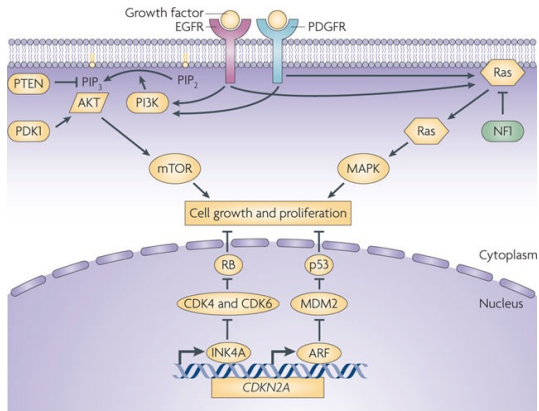
Abstract

The disruption
Parkinson's dis
antioxidant and
effects on PD.

established PD

Abstract

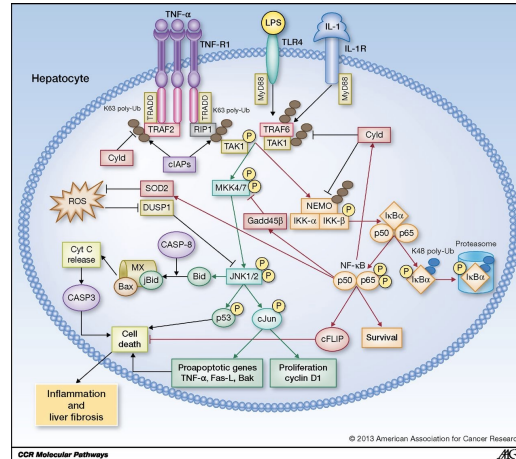
The pathological process of neurodegeneration, which is observed in Alzheimer's (AD) and Parkinson's (PD) diseases and that follows any insult to the central nervous system, is accompanied by an inflammatory reaction, which is believed to contribute to the pathogenesis of the diseases. In accordance to this, the anti-inflammatory agents are suggested to be effective in slowing or inhibiting the degenerative process. In this study, we investigated the influence of dexamethasone (DXM) on the nigrostriatal dopaminergic neurons damage following administration



Nature Reviews | Cancer



Drug targets

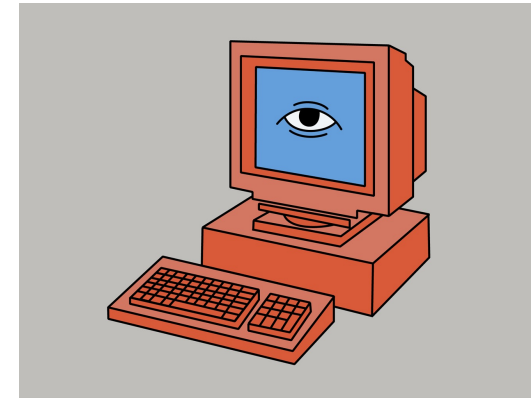


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CCR Molecular Pathways



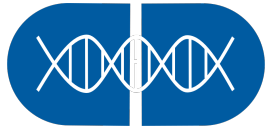
Drug targets



Drug targets

Robot-scientist





Genome Enhancer

Key result:
actionable drug targets and
prospective treatments

Enhance your genome

Disease: Colorectal Neoplasia Tissue: 1) Nothing selected (optional)

Select one or more diseases you are studying

One or more of the disease(s) you have selected can be classified as cancer. Genome Enhancer can create an additional MTB (Molecular Tumor Board) report on your genomics patient data. Please tick the checkbox below if you would like to generate the MTB report. You can refer to the user guide for further info.

Generate MTB report in addition to standard Genome Enhancer analysis report

Choose your data for analysis:

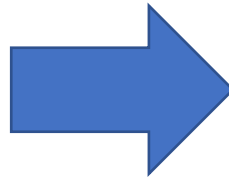
Select the conditions you want to compare in your analysis. If only one condition will be analyzed, remove the unnecessary Condition 2 by clicking on Less conditions button. If two or more conditions will be analyzed, specify the condition which refers to the baseline (control/background set).

Condition 1: Experiment: short-term survival

[+ More conditions](#) [- Less conditions](#)

[Start analysis](#)

Progress bar: Select project, Upload my data, Prepare my data, Describe my data, **Start analysis**, See report



Sequence and Pathway analysis Identification of master-regulators in gene regulatory and signal transduction pathways.

Your name
Your organization
Data received on 01/06/2017; Run on 01/06/2017; Report generated on 01/06/2017



Summary

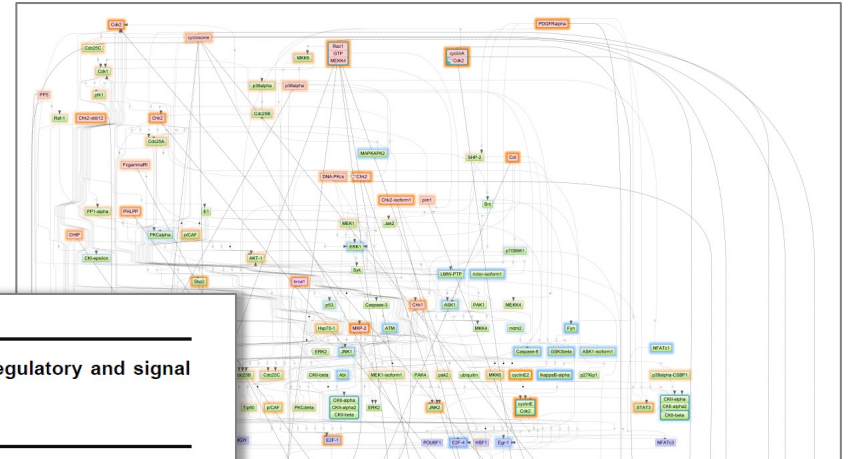
In this report we present the results of causal multi-omics data analysis, which was performed by the automatized pipeline system "From genome to target". The goal of the pipeline is to identify master regulators in gene regulatory networks as potential drug targets for the studied pathological process. On the first step of analysis we discover transcription factors that regulate genes in pathological state. The second step of analysis performs the search for so-called master-regulators, which control transcription factors that were found on the first step. The identified master-regulators are potential targets for the studied disease. After the druggability checkup, the most promising master-regulators are chosen as potential drug targets for the analyzed pathology.

Here we applied the pipeline "From genome to target" for analysis of multi-omics data set that contains transcriptomics data in **carcinoma**. The results of this analysis helped us to better understand the molecular mechanisms of the studied pathological state of the disease. Such approach promises to be very effective for rapid and accurate identification of disease drug targets with true potential.

Introduction

Multiple "omics" data are generated worldwide measuring gene and protein expression, identifying genetic and epigenetic changes and discovering disease causing mutations and variations for various pathological states of multiple organisms. Still the challenge remains to reveal deep molecular mechanisms underlying the various changes in omics data collected from the pathological states in comparison to the norm. The causal molecular mechanisms of diseases on the level of cellular regulatory networks can be described by specific pathological epigenetic changes in genomes. The molecular regulatory networks of cells are being rewired in disease conditions and such rewirings eventually lead to pathology progression. Reconstruction of the disease-specific regulatory networks and identification of potential master regulators of such networks can give us a clue on what potential ways of blocking the pathological regulatory cascades exist. Suppression of certain molecular targets can eventually let us stop the pathological process and cure the disease. Common approaches of statistical omics data analysis cannot reconstruct the cell regulatory networks due to the inability of detection of complex signal hierarchy. Thus such approaches provide only a very limited clue to the causes of the observed phenomena and actually do not lead to the understanding of the pathology molecular mechanism.

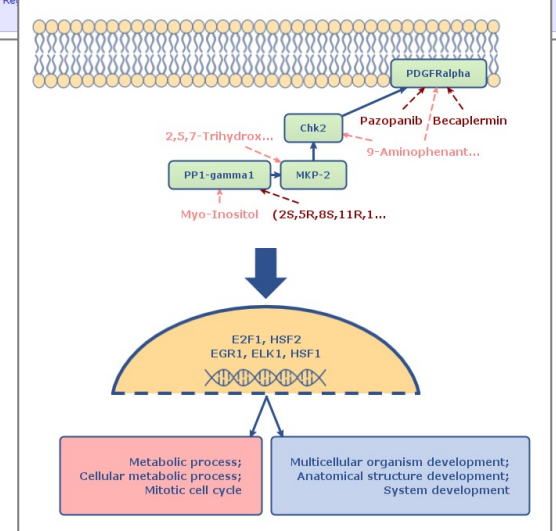
Unlike common approaches, the "upstream analysis" method [1-5], integrated in the pipeline system "From genome to target", performs causal interpretation of the observed changes in the pathology state. This approach comprises two major steps: (1) analysis of enhancers of identified differentially expressed genes (DEGs) to reveal transcription factors (TFs) involved in the process under study; (2) reconstruction of signaling pathways that activate these TFs and identification of master-regulators on the top of such pathways. The first step is done with the help of the TFs binding sites database and site identification algorithms - Match [6] and CMA [7]. The second step is done with the help of intracellular signal transduction database and special graph search algorithms implemented in the pipeline system "From genome to target". The upstream analysis approach is integrated in the pipeline system with certain improvements, such as dynamical simulation of the constructed signal transduction network and druggability check of the revealed targets (with the use of QJSAR approaches). Therefore the applied in this study pipeline system "From genome to target" opens new perspectives to process various omics data by complete automatization of such complex tasks as disease molecular mechanism identification and drug target selection.



Drugs approved in clinical trials

Table 14. FDA approved drugs or drugs used in clinical trials for the studied pathology (most promising treatment candidates selected for the identified drug targets on the basis of literature curation in HumanPSD™ database)
See full table →

Name	Target names	Drug rank	Disease activity score	Phase 4	Status (provided by Drugbank)
Pazopanib	ITK, KDR, FLT1, PDGFRB, PDGFRA	7	7	Carcinoma, Renal Cell, Neoplasms, Noma	small molecule, approved
Sunitinib	KDR, PDGFRB, FLT1, PDGFRA	36	2	Carcinoma, Renal Cell, Gastrointestinal Neoplasms, Gastrointestinal Stromal Tumors, Intestinal Neoplasms, Lung Neoplasms, Neoplasms, Neuroendocrine Tumors...	small molecule, approved, investigational
Regorafenib	KDR, FLT1,			Colorectal Neoplasms, Gastrointestinal Stromal	



Thank you!



Funding

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Russia



ФЦП

Исследования и разработки