



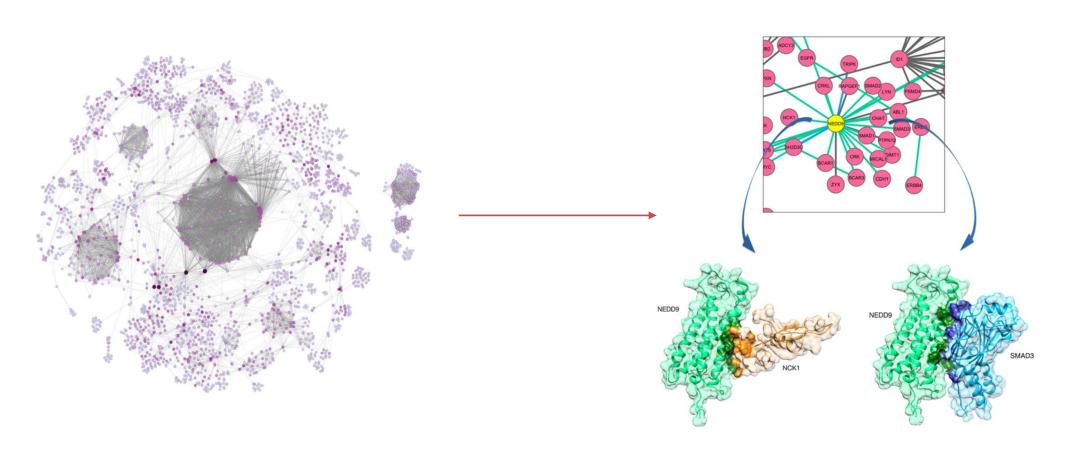
Prediction of the Function of Protein Interactions in Post-Translational Modifications

2023 M.Sc. Biology & Biomedicine

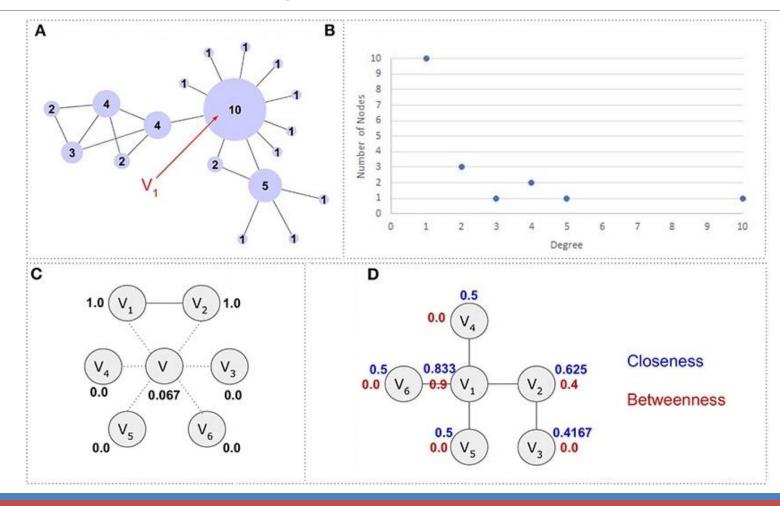
Module: Proteinbiochemie und Bioinformatik

Aimilia Christina Vagiona PhD student Supervisor: Miguel Andrade

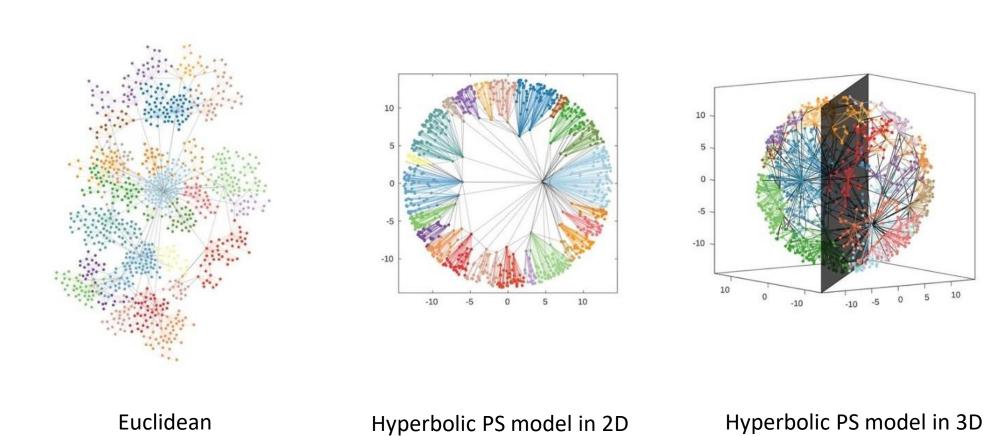
Why do we study the human interactome?



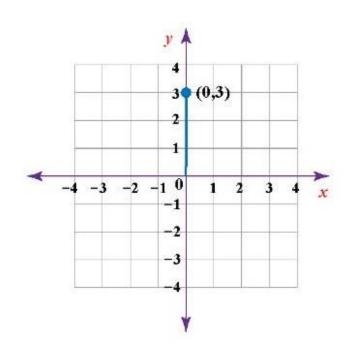
Network theory

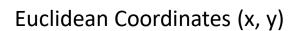


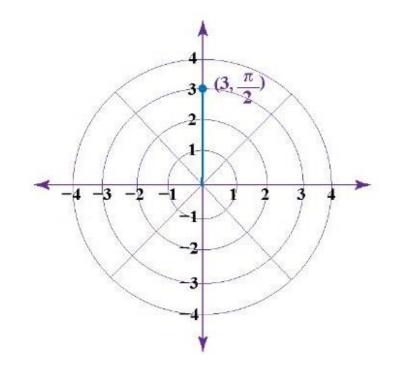
Hyperbolic network model



PS model



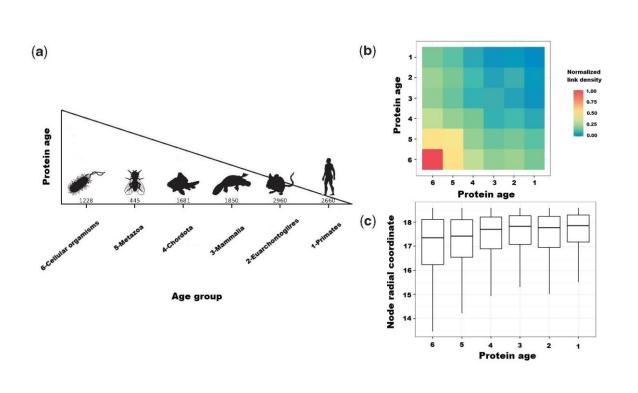


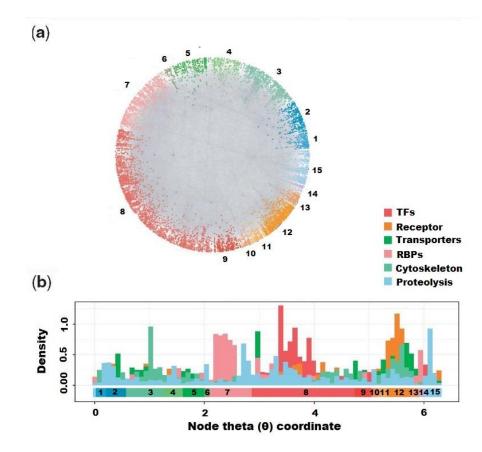


Polar Coordinates (r, theta)

- 1) Node Sorting: Nodes are sorted in decreasing order by degree
- 2) Starting Point: The first node is placed at the center and it gets a random angle
- 3) Expanding the Space: embedding new nodes:
- Distance from Center: in the new node assigned radial coordinate *ri* = 2ln*i*
- Adjusting Positions: The positions of existing nodes are adjusted based on their distances from the new node ($r_j(i) = \beta r_j + (1 \beta) r_i$)
- Finding the Best Angle: The new node is given an anglular coordinate that maximizes the likelihood of the PS model
- 4) Repeat for All Nodes: Steps 2 and 3 are repeated for every node in the network.

hPIN in the Hyperbolic Space





hPIN in the Hyperbolic Space

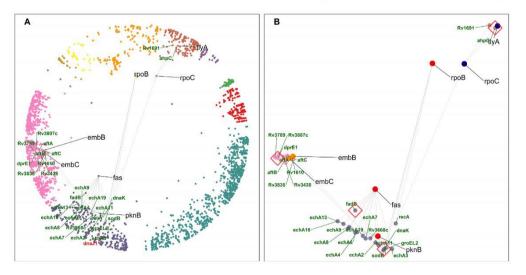


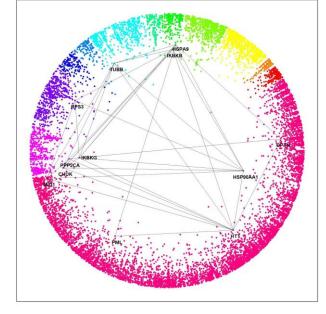


Article

Selection of Multi-Drug Targets against Drug-Resistant Mycobacterium tuberculosis XDR1219 Using the Hyperbolic Mapping of the Protein Interaction Network

Noor ul Ain Zahra ^{1,2}, Aimilia-Christina Vagiona ², Reaz Uddin ^{1,*} o and Miguel A. Andrade-Navarro ^{2,*} o





International Journal of Molecular Sciences

MDPI

Article

Analysis of Huntington's Disease Modifiers Using the Hyperbolic Mapping of the Protein Interaction Network

Aimilia-Christina Vagiona ¹, Pablo Mier ¹, Spyros Petrakis ² and Miguel A. Andrade-Navarro ^{1,*}

Hyperbolic applications

nature > scientific reports > articles > article

Article Open access Published: 06 August 2021

The inherent community structure of hyperbolic networks

Scientific Reports 11, Article number: 16050 (2

nature > nature communications > articles > article

Article Open access Published: 05 May 2021

Deep generative model embedding of single-cell RNA-Seq profiles on hyperspheres and hyperbolic spaces

10023

Jiarui Ding ≥ & Aviv Regev

Nature Communications 12, Article number: 2554 (2021) | Cite this article

IEEE TRANSACTIONS ON PATTERN ANALYSIS AND MACHINE INTELLIGENCE, VOL. 44, NO. 12, DECEMBER 2022

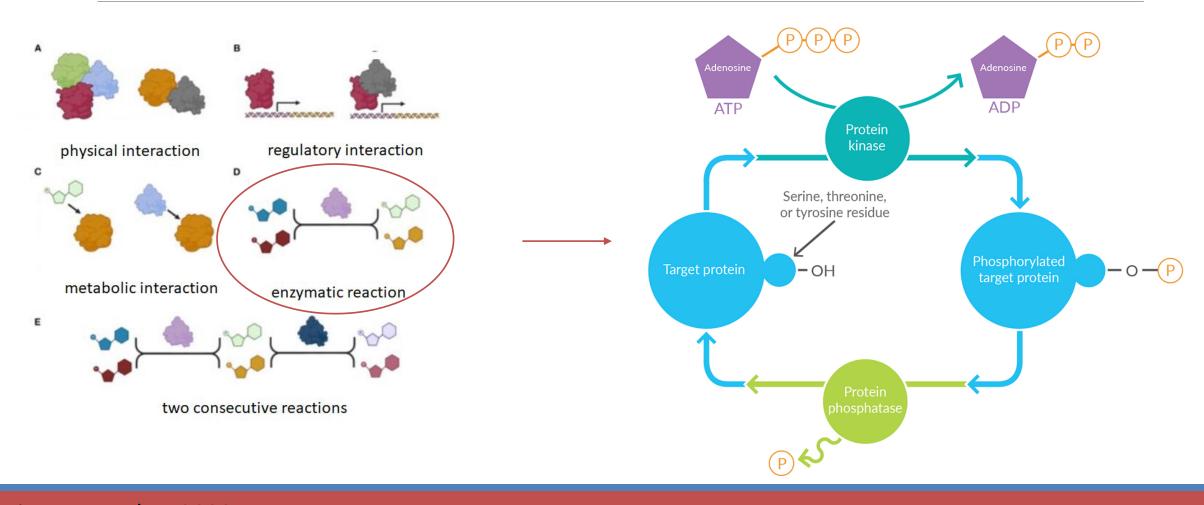
Hyperbolic Deep Neural Networks: A Survey

Wei Peng[®], Tuomas Varanka, Abdelrahman Mostafa, Henglin Shi[®], and Guoying Zhao[®], *Fellow, IEEE*

Abstract—Recently, hyperbolic deep neural networks (HDNNs) have been gaining momentum as the deep representations in the hyperbolic space provide high fidelity embeddings with few dimensions, especially for data possessing hierarchical structure. Such a hyperbolic neural architecture is quickly extended to different scientific fields, including natural language processing, single-cell RNA-sequence analysis, graph embedding, financial analysis, and computer vision. The promising results demonstrate its superior capability, significant compactness of the model, and a substantially better physical interpretability than its counterpart in the euclidean space. To stimulate future research, this paper presents a comprehensive review of the literature around the neural components in the construction of HDNN, as well as the generalization of the leading deep approaches to the hyperbolic space. It also presents current applications of various tasks, together with insightful observations and identifying open questions and promising future directions.

Index Terms—Neural networks on Riemannian manifold, hyperbolic neural networks, Poincaré model, Lorentz model

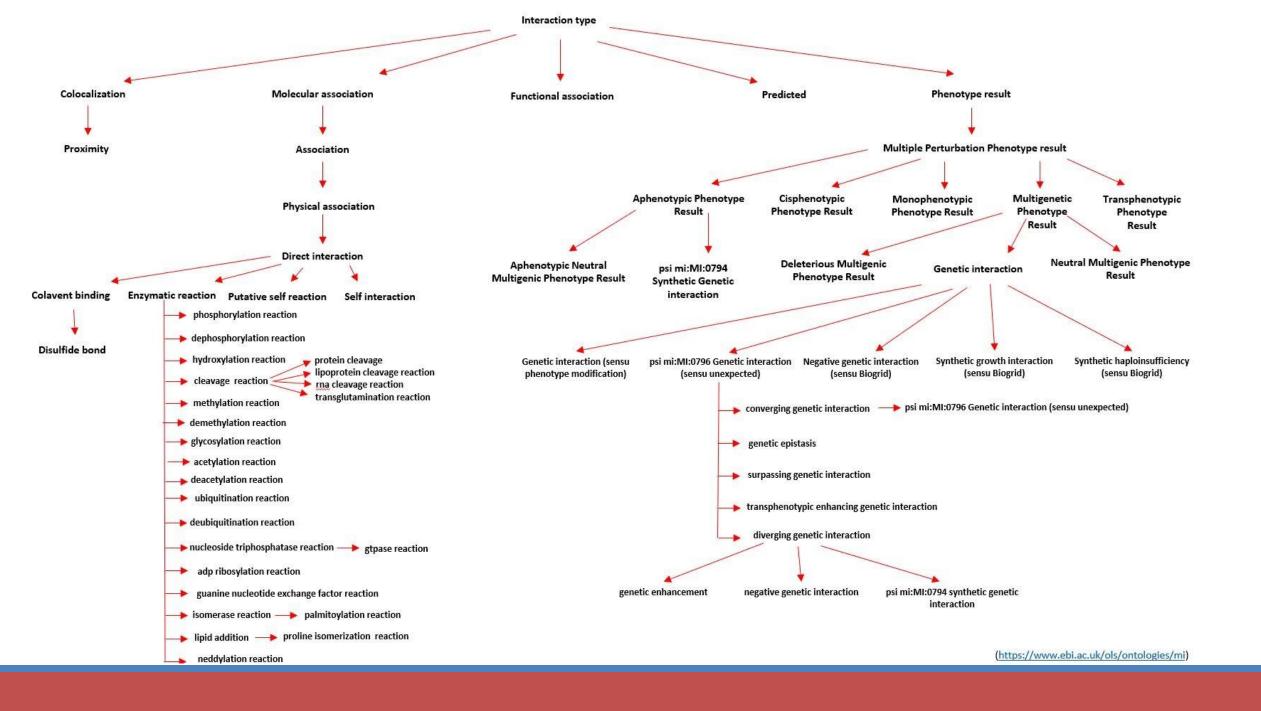
Function of PPIs



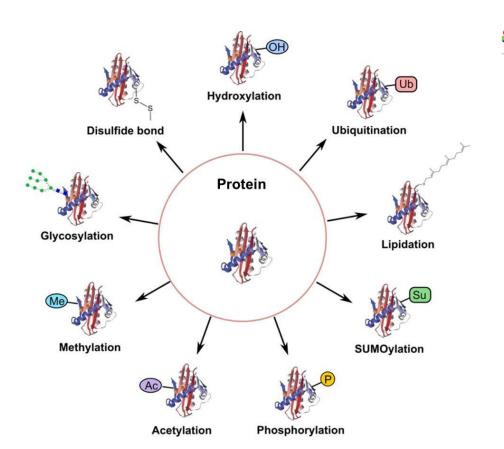
Molecular annotations of PPIs

The Molecular Interactions (MI) ontology forms a structured controlled vocabulary for the annotation of experiments concerned with protein-protein interactions.





Why do we focus on PTMs?





REVIEW cublished: 19 September 2018 doi: 10.3389/Incel.2018.00290



Roles of Post-translational Modifications in Spinocerebellar Ataxias

Linlin Wan¹⁷, Keqin Xu¹⁷, Zhao Chen¹, Beisha Tang^{1,2,3,4,5,6,7} and Hong Jiang^{1,2,3,4,8}*

¹ Department of Neurology, Narroge i-Ropalia, Contral South Linkwestly, Changdina, China, ³ Mallorius Clinical Risosarch Contents of Gradientic Diseases, Contral South Linkwestly, Chinage (Linia, ³ Reyl Jacknotory of Human Province in Navarchapmentel Victoria, Carlorial South Linkwestly, Chinage (18 April 18 April 18





Remier

Do Post-Translational Modifications Influence Protein Aggregation in Neurodegenerative Diseases: A Systematic Review

Larissa-Nele Schaffert and Wayne G. Carter *0

School of Medicine, University of Nottingham, Royal Derby Hospital Centre, Uttoxeter Road, Derby DE22 3DT, UK; larissaschaffert@yahoo.de

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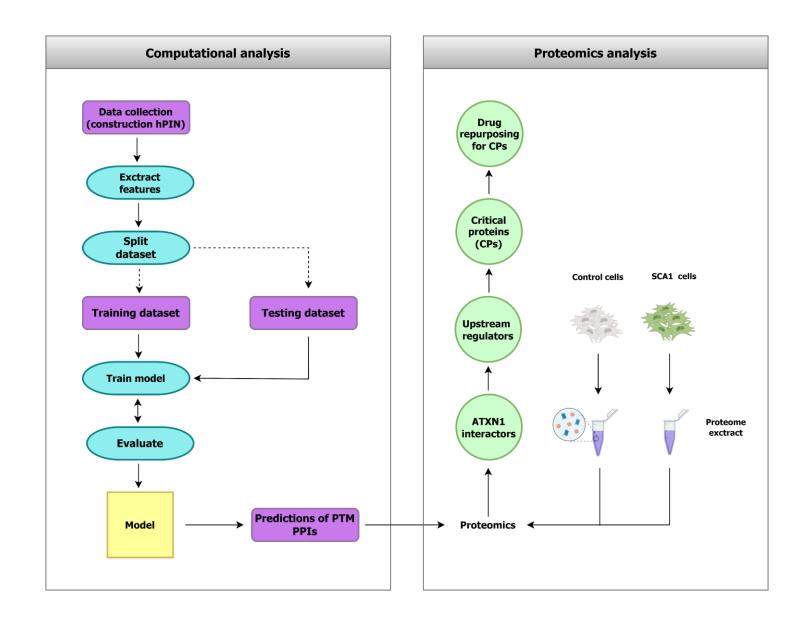
Review

Post-translational modifications: Regulators of neurodegenerative proteinopathies

Rohan Gupta 1, Mehar Sahu 1, Devesh Srivastava 1, Swati Tiwari 1, Rashmi K. Ambasta, Pravir Kumar *

Molecular Neuroscience and Functional Genomics Laboratory, Department of Biotechnology, Delhi Technological University (Formerly DCE), Delhi 110042, India

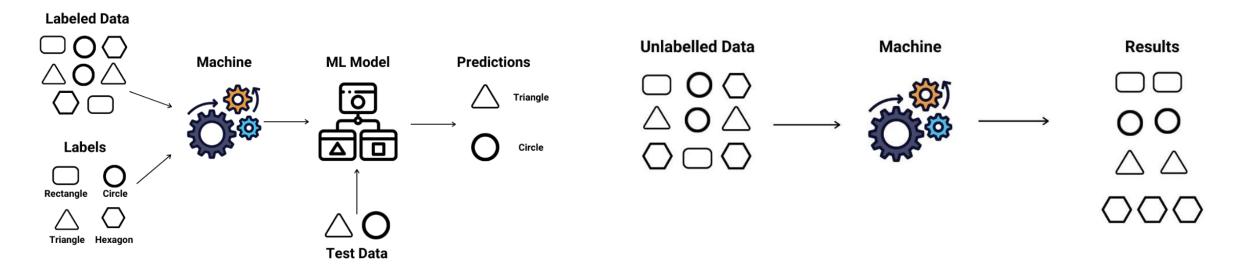


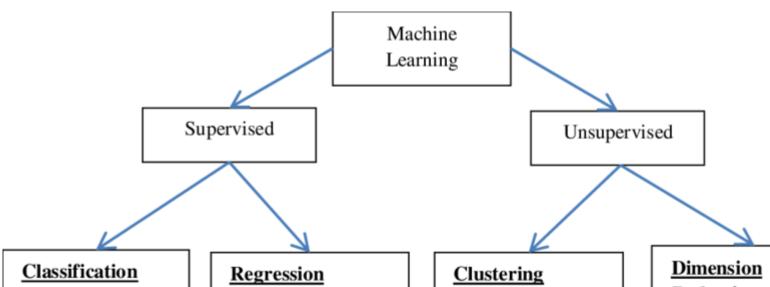


What is machine learning?

Supervised Learning

Unsupervised Learning





Predicting a categorical variable

Input: Labeled

data set

Output: Discrete

values

Algorithms:

Decision Trees Support Vector Machines

Predicting a numeric variable

Input: Labeled data

set

Output-Continuous

Values

Algorithms

Linear Regression

Decision Trees

Random forests

Identify a pattern or groups of similar objects

Algorithms

K-Means Clustering ANN (Artificial Neural Networks)

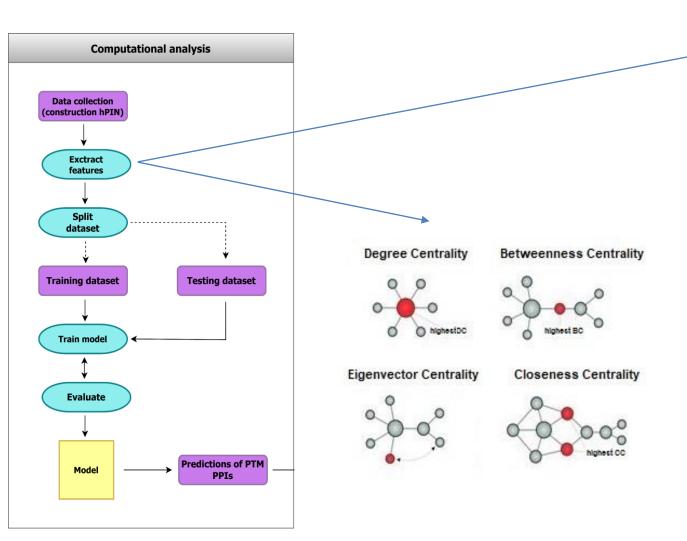
Reduction

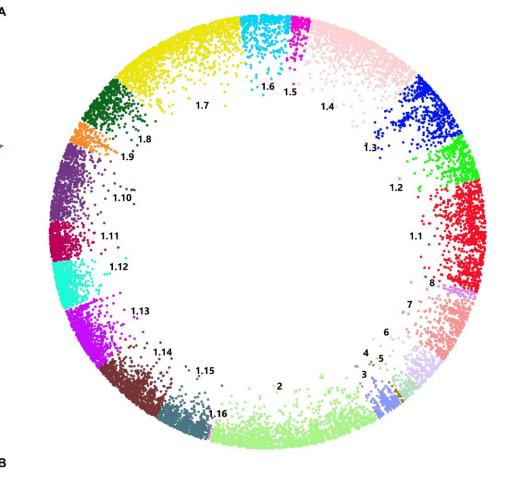
Reduces the number of variables being considered to the find exact information required

Algorithms

Principal component Analysis (PCOA)

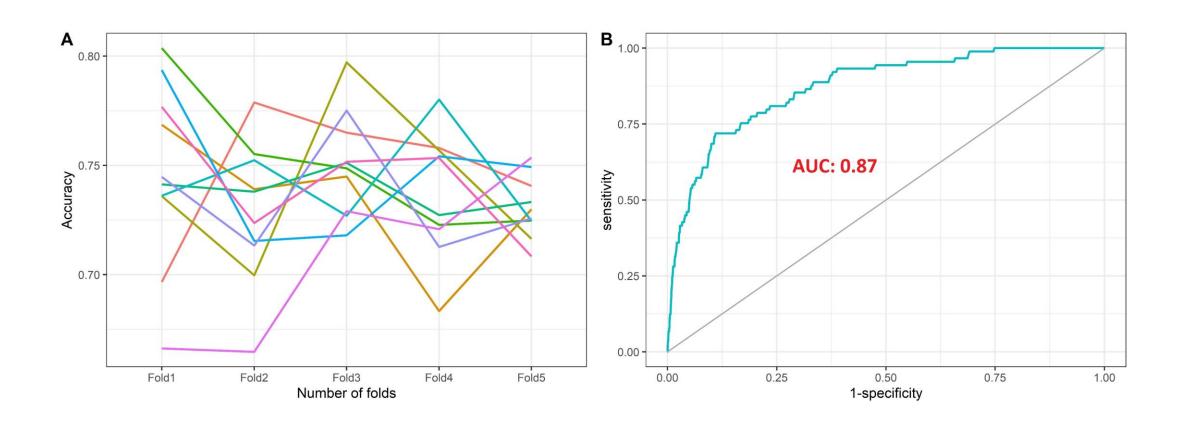
Features



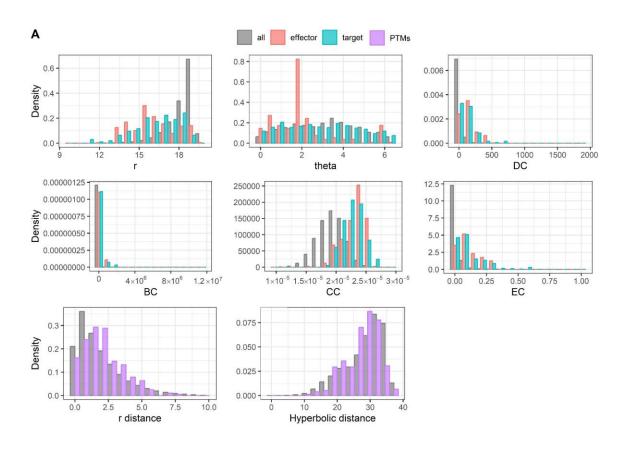


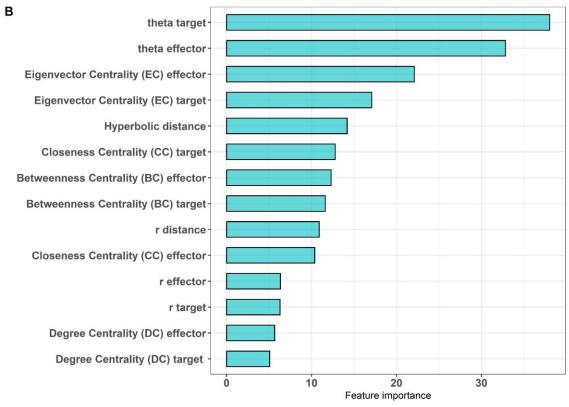
Cluster	GO BP Terms	Cluster	GO BP Terms
1.1	mRNA processing	1.13	mitochondrial translational elongation
1.2	regulation of gene silencing by miRNA	1.14	mitochondrial respiratory chain complex assembly
1.3	regulation of nucleobase-containing compound metabolic process	1.15	sphingolipid metabolic process
1.4	alpha-amino acid biosynthetic process	1.16	regulation of angiotensin levels in blood
1.5	cell cycle G1/S phase transition	2	regulation of cell cycle G2/M phase transition
1.6	post-translational protein modification	3	intraciliary retrograde transport
1.7	protein ubiquitination	4	N-terminal protein amino acid modification
1.8	positive regulation of establishment of protein localization to telomere	5	positive regulation of tyrosine phosphorylation of STAT protein
1.9	regulation of cellular amine metabolic process	6	positive regulation of lipopolysaccharide-mediated signaling pathway
1.10	metanephric renal vesicle morphogenesis	7	chemokine-mediated signaling pathway
1.11	protein transport	8	regulation of DNA-templated transcription, initiation
1.12	protein insertion into ER membrane by stop-transfer membrane-anchor sequence		

RF model

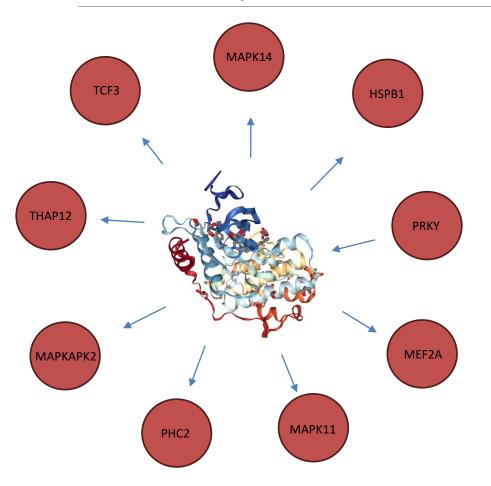


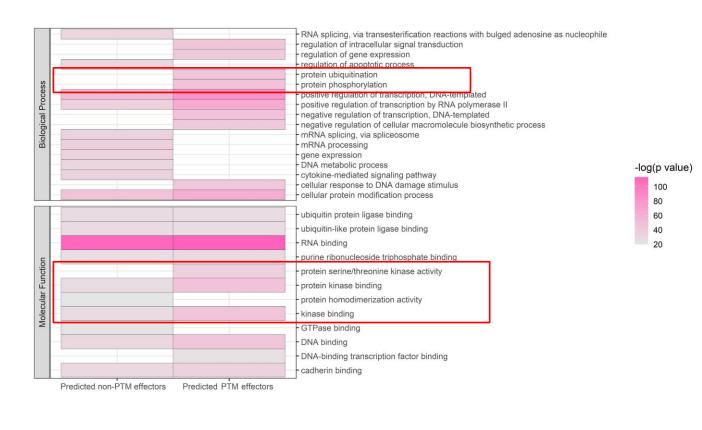
Feature importance



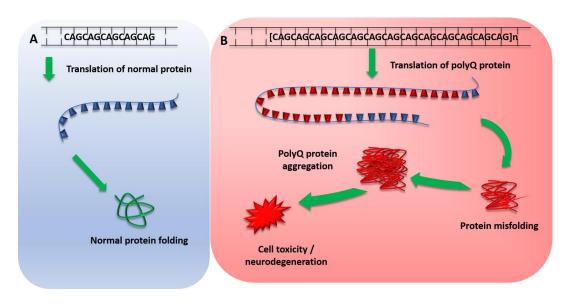


MAPK3 protein kinase

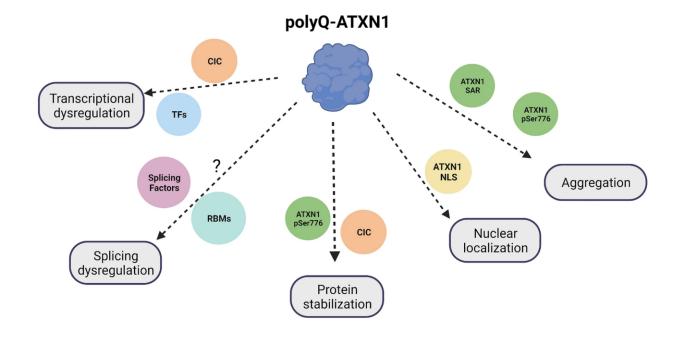




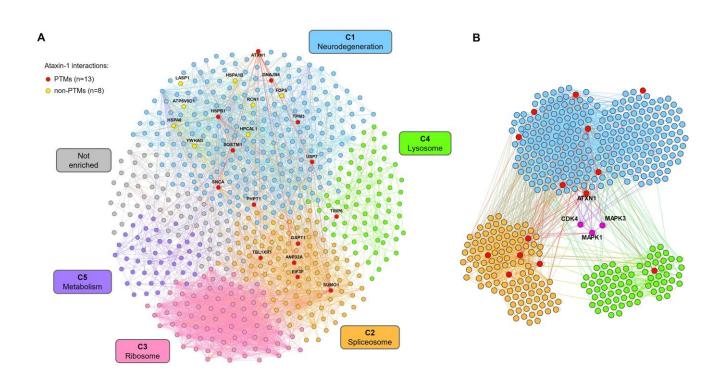
Spinocerebellar ataxia 1 (SCA1)

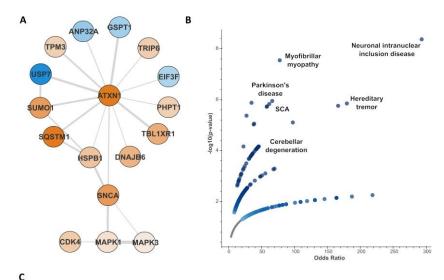


Molecular mechanisms of SCA1 pathogenesis



SCA1 cell model- Predicted PTMs





Drug candidate	Reverse score	Administration	
artesunate		Malaria	
BRD-K54687541	25%	Unknown	
betamethasone	25%	cerebral edema, congenital adrenal hyperplasia	
budesonide	25%	Crohn's disease, asthma, allergies	
BRD-K71265179	25%	Unknown	
Linifanib	25%	hematologic malignancies, solid tumors	
BRD-A08662020	25%	Unknown	

Next steps

Experimental validation of the predicted directed PTMs

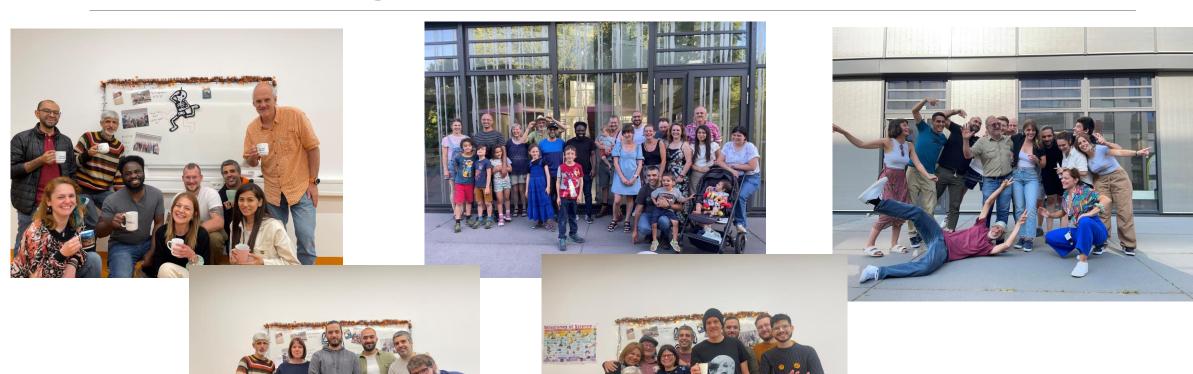
Experimental verification of the suggested drugs

Explore hyperbolic geometry in different biological data

References

- Gregorio Alanis-Lobato, Pablo Mier, Miguel Andrade-Navarro, The latent geometry of the human protein interaction network, Bioinformatics, Volume 34, Issue 16, August 2018, Pages 2826–2834
- Laidou S, Alanis-Lobato G, Pribyl J, et al. Nuclear inclusions of pathogenic ataxin-1 induce oxidative stress and perturb the protein synthesis machinery
- Zahra NuA, Vagiona AC, Uddin R, Andrade-Navarro MA. Selection of Multi-Drug Targets against Drug-Resistant Mycobacterium tuberculosis XDR1219 Using the Hyperbolic Mapping of the Protein Interaction Network. Int. J. Mol. Sci. 2023, 24, 14050
- Vagiona AC, Mier P, Petrakis S, Andrade-Navarro MA. Analysis of Huntington's Disease Modifiers Using the Hyperbolic Mapping of the Protein Interaction Network. Int J Mol Sci. 2022;23(10):5853
- •Sambataro F, Pennuto M. Post-translational Modifications and Protein Quality Control in Motor Neuron and Polyglutamine Diseases. Front Mol Neurosci. 2017;10:82

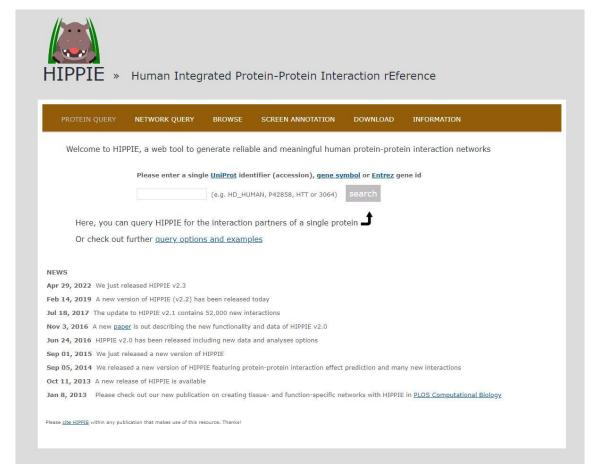
Acknowledgments



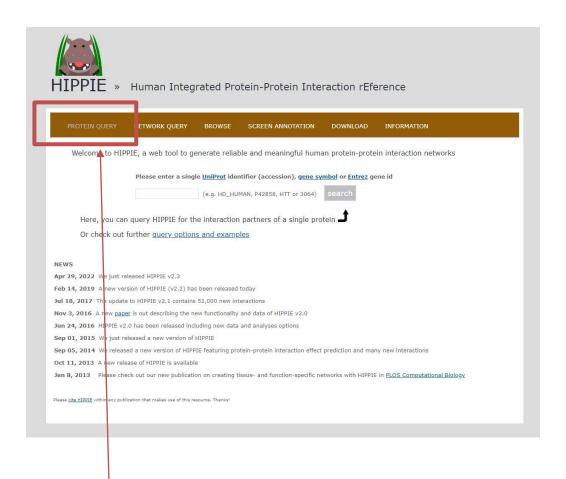
Lovely to meet you all !!!!!

Thank you!
Any questions?

Let's play...



http://cbdm-01.zdv.uni-mainz.de/~mschaefer/hippie/

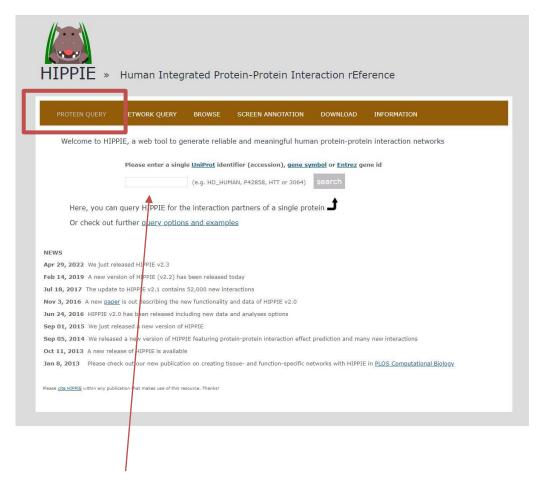


Input: one protein

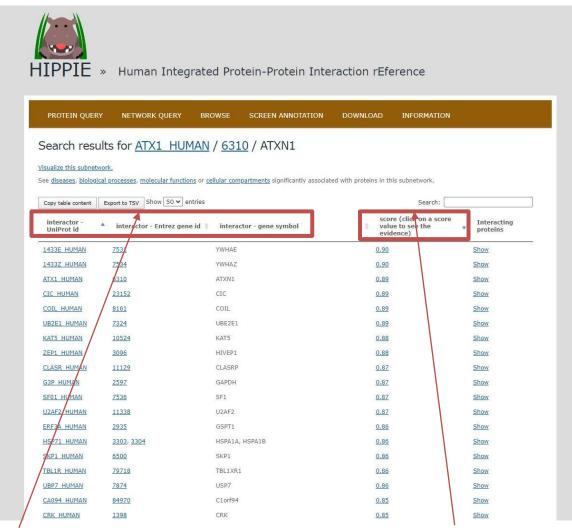


Input: list of protein

How many interactors ATXN1 has?



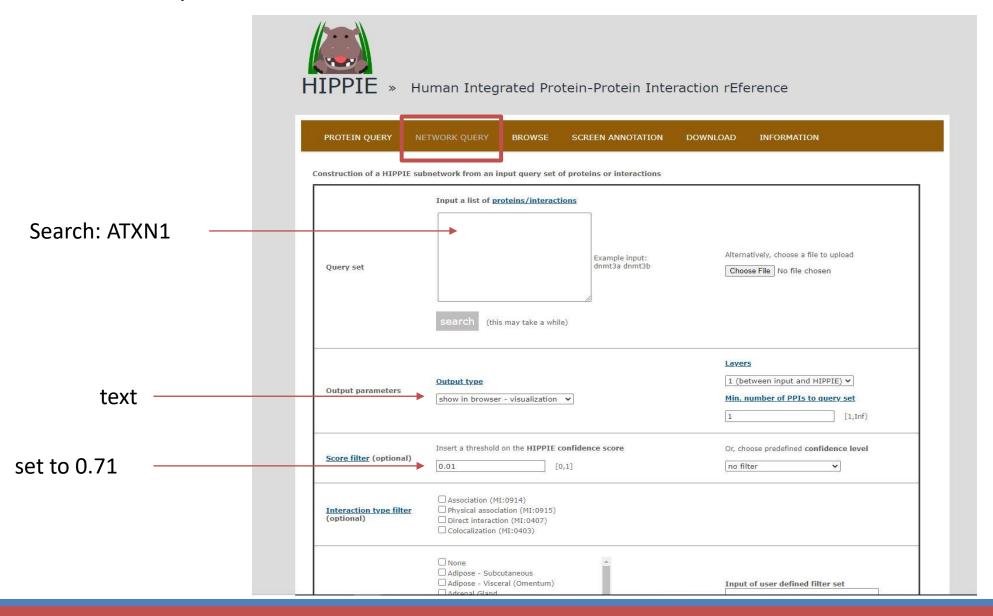
Search: ATXN1



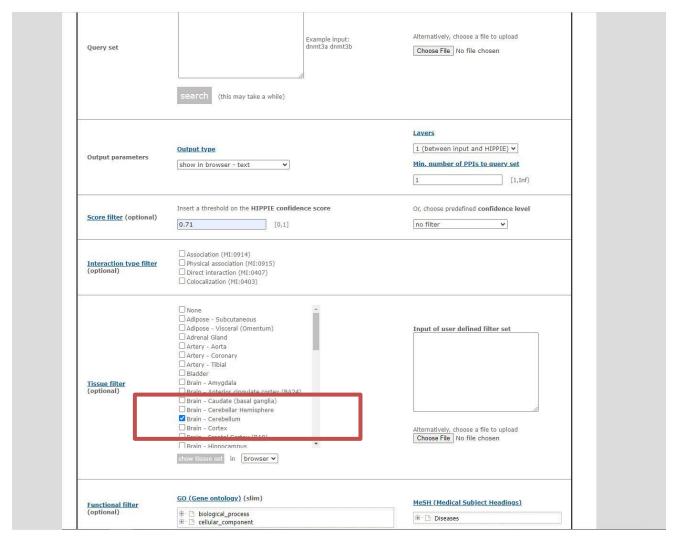
Interactors of ATXN1

confidence score

How many interactors ATXN1 has with confidence score more than 0.71?



How many interactors ATXN1 has with confidence score more than 0.71 that are expressed in the cerebellum?



How many interactors ATXN1 has, with confidence score more than 0.71 that are expressed in the cerebellum and associated with neurodegenerative disease?

