



# A NOVEL APPROACH TO EVALUATE MULTI-OMIC MULTIVARIATE DRUG TARGET IDENTIFICATION ALGORITHMS

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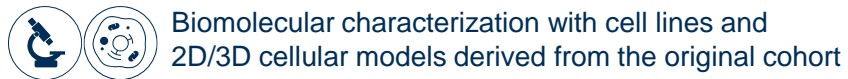
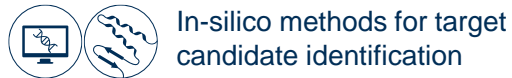
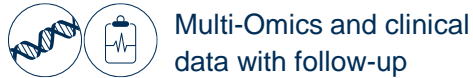
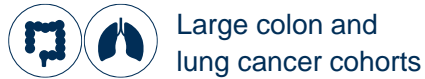
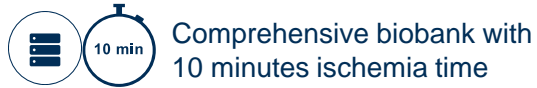
# AGENDA

1. Introduction
2. An Input Data-Driven AMI Performance Measure
3. From Single-Omics to Multi-Omics AMI
4. Conclusions and Outlook



# INTRODUCTION

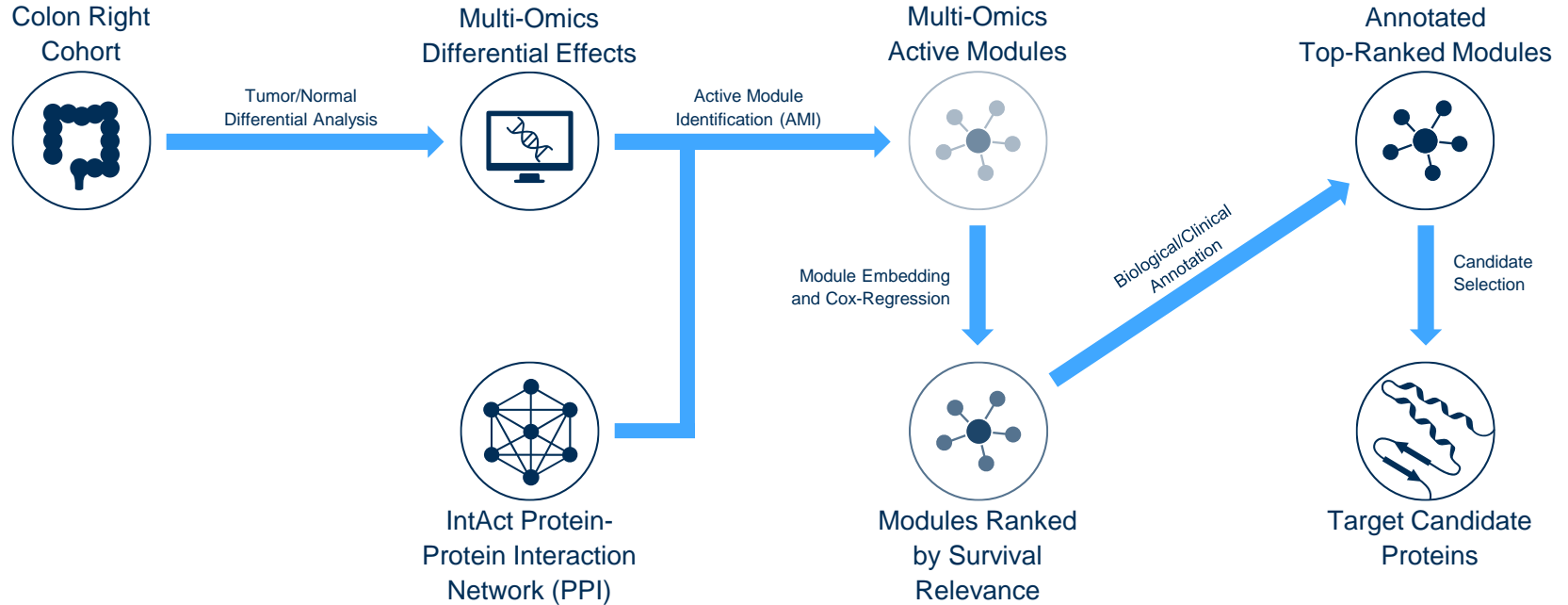
# COMPANY PROFILE



**Indivumed Therapeutics is a Hamburg-based biotech company founded in 2002 with over 80 employees.**

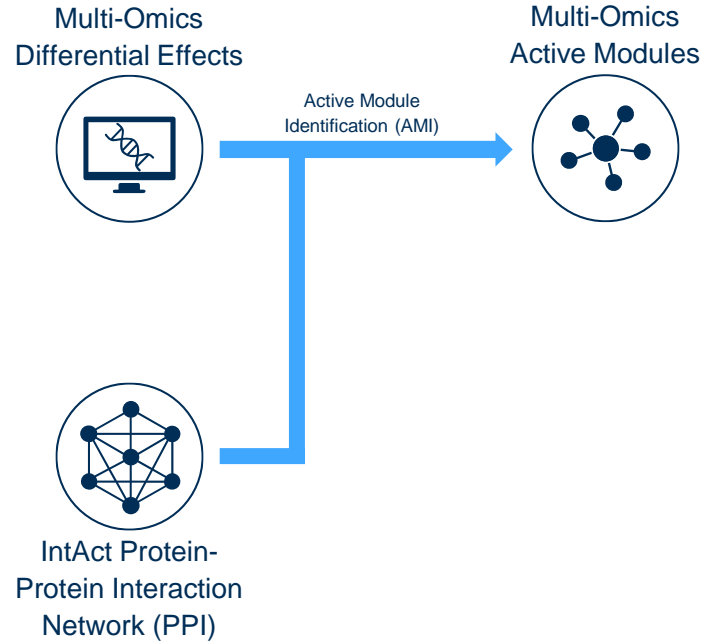
**We focus on cancer drug target identification and aim to create target candidate packages, which are screening-ready and have been validated extensively.**

# OUR APPROACH TO TARGET CANDIDATE IDENTIFICATION



# CHALLENGES IN ACTIVE MODULE IDENTIFICATION

- Performance evaluation of AMI methods rely on external biological datasets.
- Single-Omics AMI methods are abundant, but Multi-Omics AMI methods are rare.



The background is a solid dark blue color. Overlaid on this are several thin, white, stylized lines that resemble a circuit board or a network diagram. These lines are interconnected at various points, with some lines ending in small white dots. The overall aesthetic is clean, modern, and technological.

# **AN INPUT DATA-DRIVEN AMI PERFORMANCE MEASURE**

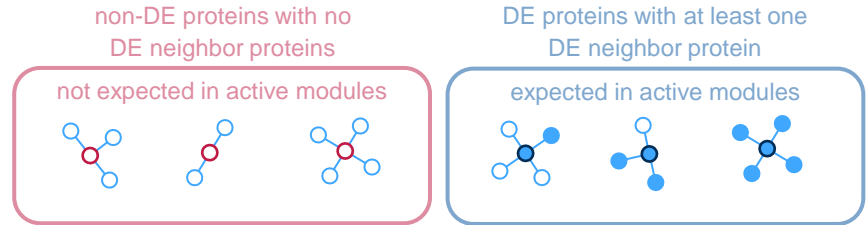


# AMI PERFORMANCE WITH ACTIVE MCC

We propose to measure AMI performance based on PPI topology and differential effects

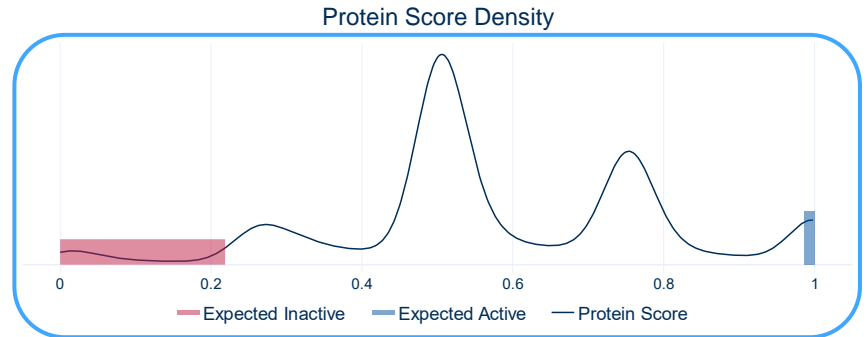
## IDEA:

Define two node classes on which AMI can be treated as node classification and measure performance with Matthew Correlation Coefficient (MCC).



## ADVANTAGES:

- Makes choice of validation dataset obsolete
- Can be used with artificial input data
- Allows for parameter tuning

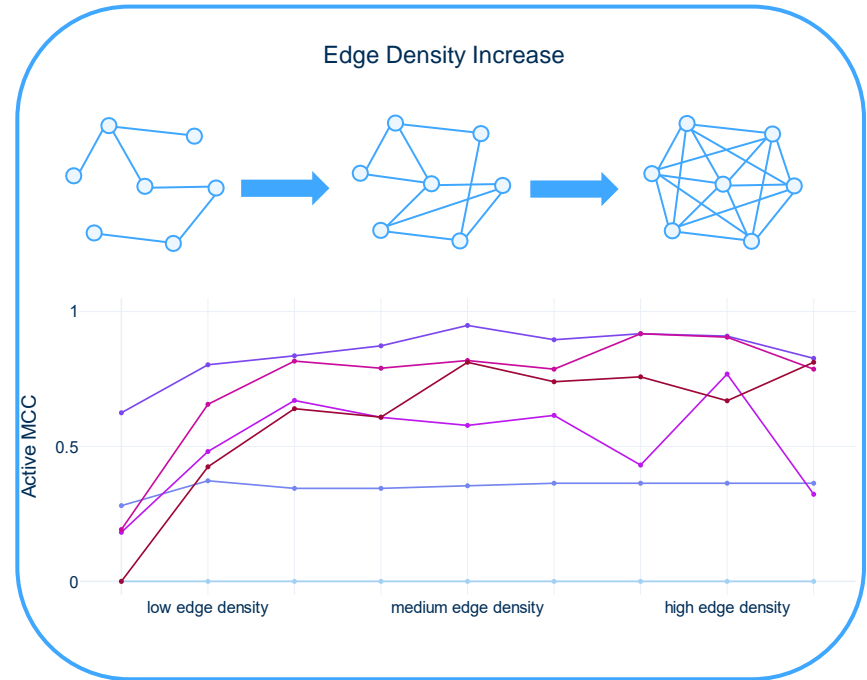
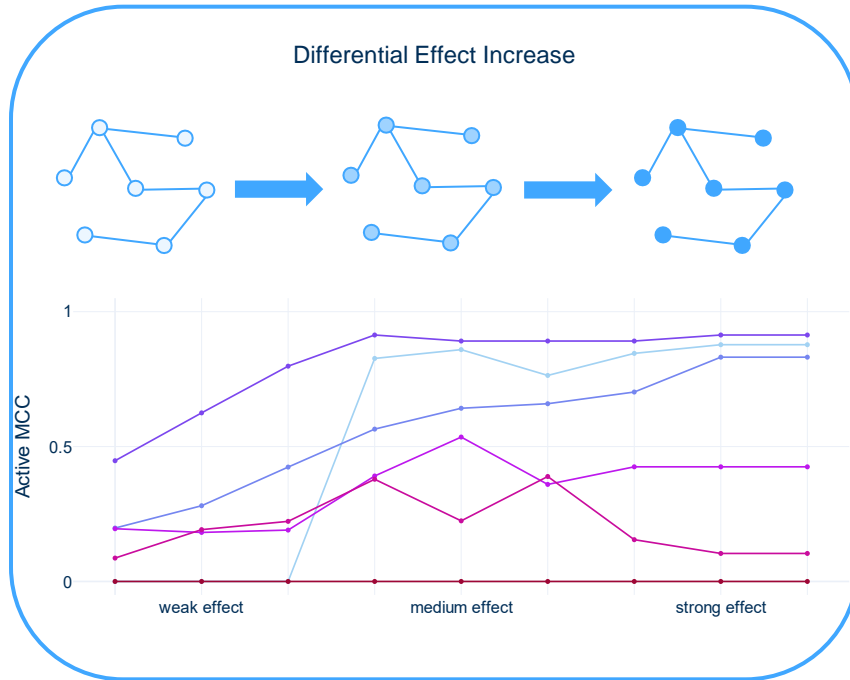






# PROOF OF CONCEPT ON ARTIFICIAL INPUT DATA

We test validity on a downsampled PPI with artificially generated disjoint active modules with varying sizes



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# FROM SINGLE-OMICS TO MULTI-OMICS AMI

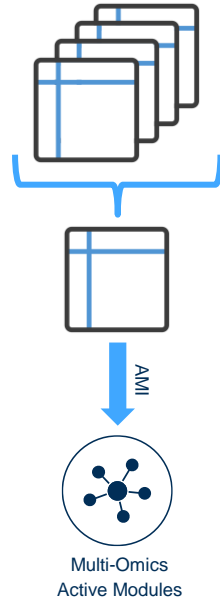


# MULTI-OMICS INTEGRATION STRATEGIES

As single-omics methods are abundant, we aim to extend them to a multi-omics setting

**Early Integration:** Aggregate multi-omics effects to single-omics effect and run single-omics method.

- Easy for any single-omics method
- Detects similar patterns across layers
- Layer specific signals might get lost

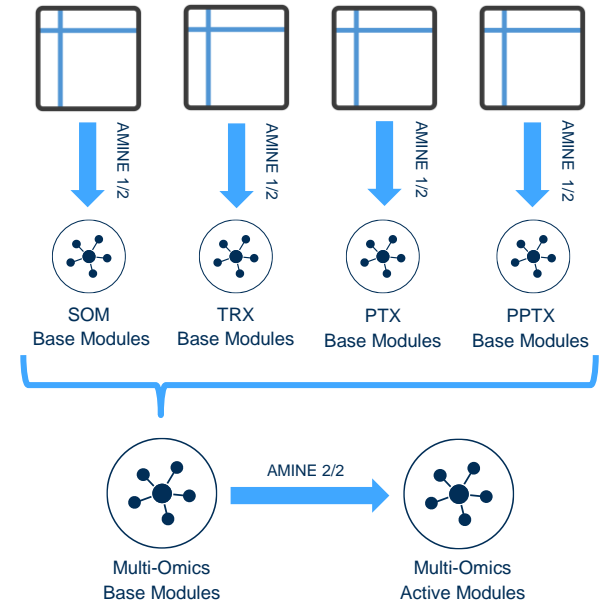


**Used:**

- RFIM ([Wang2022](#))
- DOMINO ([Levi2021](#))
- MONET ([Tomasoni2020](#))

**Middle Integration:** Extend single-omics method with an intermediate aggregation step.

- Potentially hard, depending on the single-omics method
- Detects weakly connected patterns across layers
- Layer specific signals can get through



**Extended:**

- AMINE ([Pasquier2023](#))

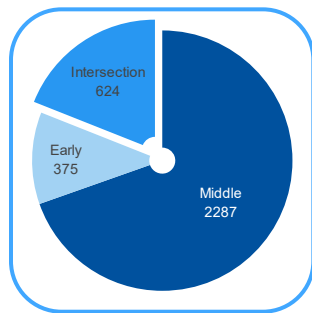


# ACTIVE MODULES FOR COLON RIGHT

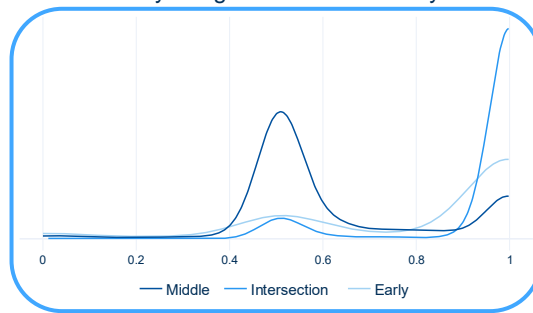
## Early vs. Middle Integration AMINE

- PPI: IntAct.246 with physical and functional interactions
- SOM: Somatic mutation overrepresentation
- TRX: Differential expression in RNA-seq layer
- PTX: Differential expression in Proteomics layer
- PPTX: Differential expression in Phosphorylation layer

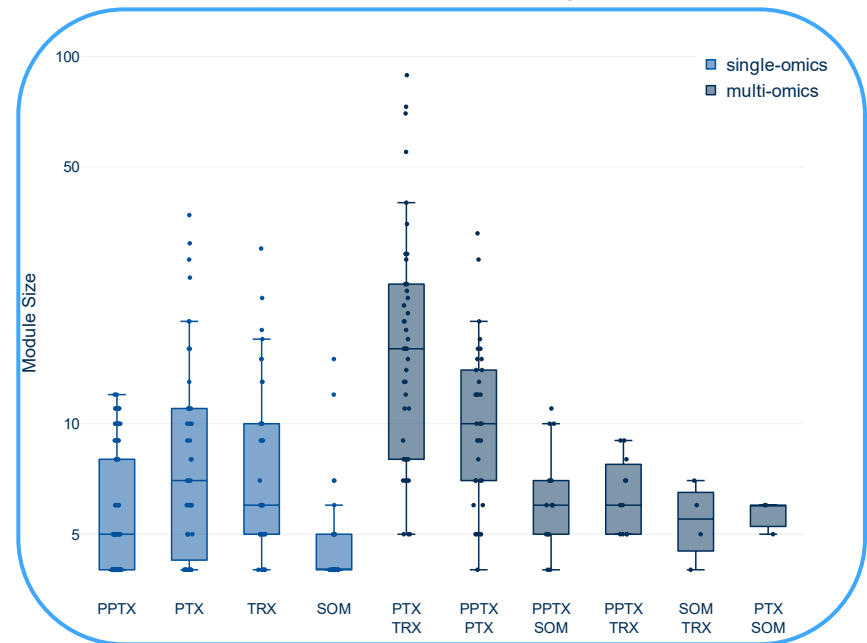
AMINE Active Proteins



Early Integration Effect Density



Module Type Statistics for Middle Integration AMINE





# CONCLUSIONS AND OUTLOOK

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- We developed an AMI performance measure derived only by input data
- We created a Multi-Omics extension of AMINE with middle integration
- We currently analyze our colon right cohort with the described multi-omic multivariate workflow
- In the future we are going to study other cancer cohorts



# REFERENCES

- **Ischemia time analysis:** Von der Heyde et al., 2024, *Tumour specimen cold ischemia time impacts molecular cancer drug target discovery*, <https://doi.org/10.1038/s41419-024-07090-x>
- **AMINE:** Pasquier et al., 2023, *A network embedding approach to identify active modules in biological interaction networks*, <https://doi.org/10.26508/lsa.202201550>
- **RFIM:** Wang et al., 2022, *A statistical physics approach for disease module detection*, <https://doi.org/10.1101/2Fgr.276690.122>
- **DOMINO:** Levi et al., 2021, *DOMINO: a network-based active module identification algorithm with reduced rate of false calls*, <https://doi.org/10.15252/msb.20209593>
- **MONET:** Tomasoni et al., 2020, *MONET: a toolbox integrating top-performing methods for network modularisation*, <https://doi.org/10.1093/bioinformatics/btaa236>
- **Disease Module Detection DREAM challenge meta-analysis:** Choobdar et al., 2019, *Assessment of network module identification across complex diseases*, <https://doi.org/10.1038/s41592-019-0509-5>

# THANK YOU!

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