

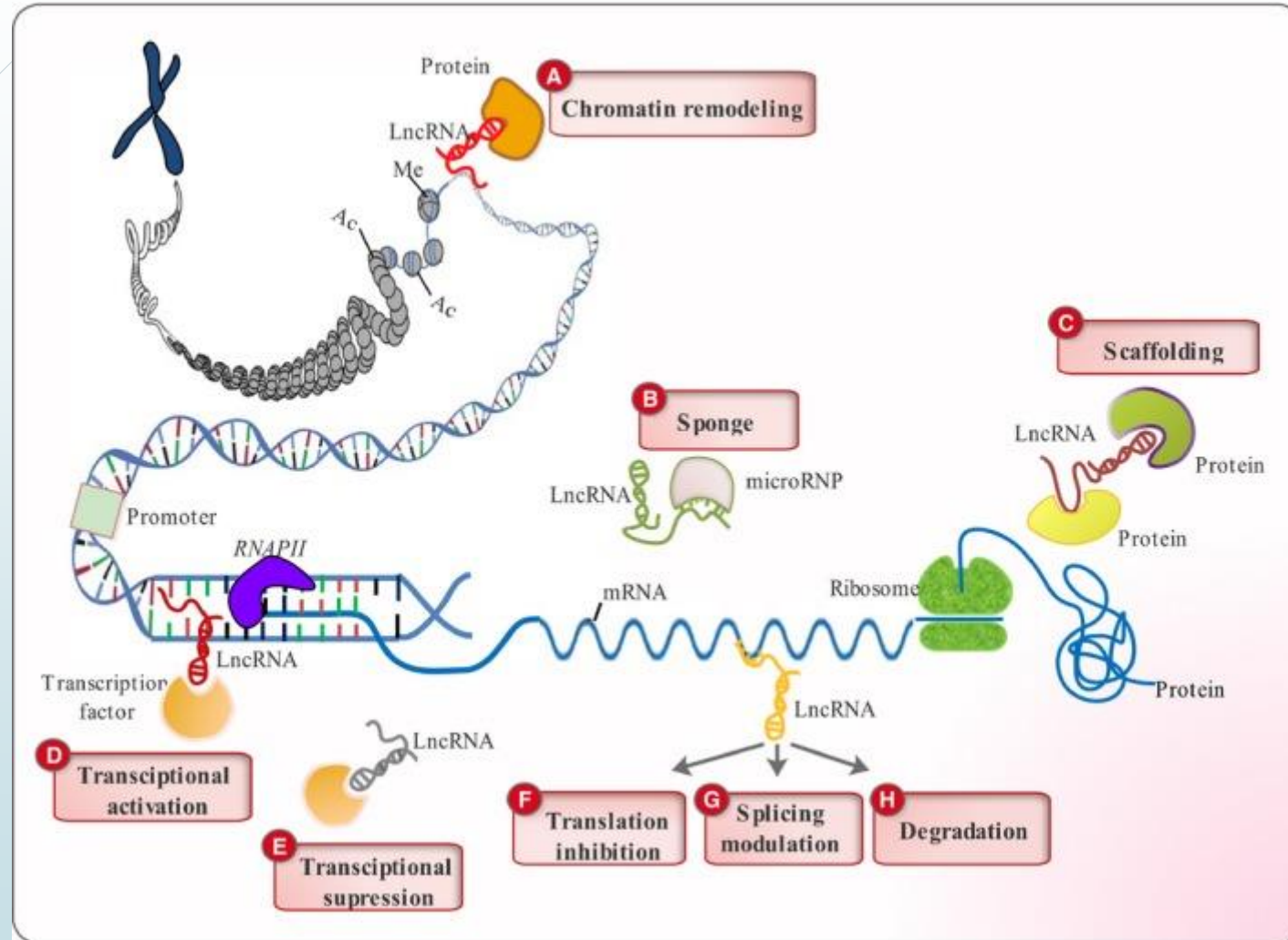


Neighborhood based approaches for the prediction of lncRNA-Disease association from tripartite graphs

Ricercatori coinvolti: **Mariella Bonomo**, Armando La Placa, Simona E. Rombo

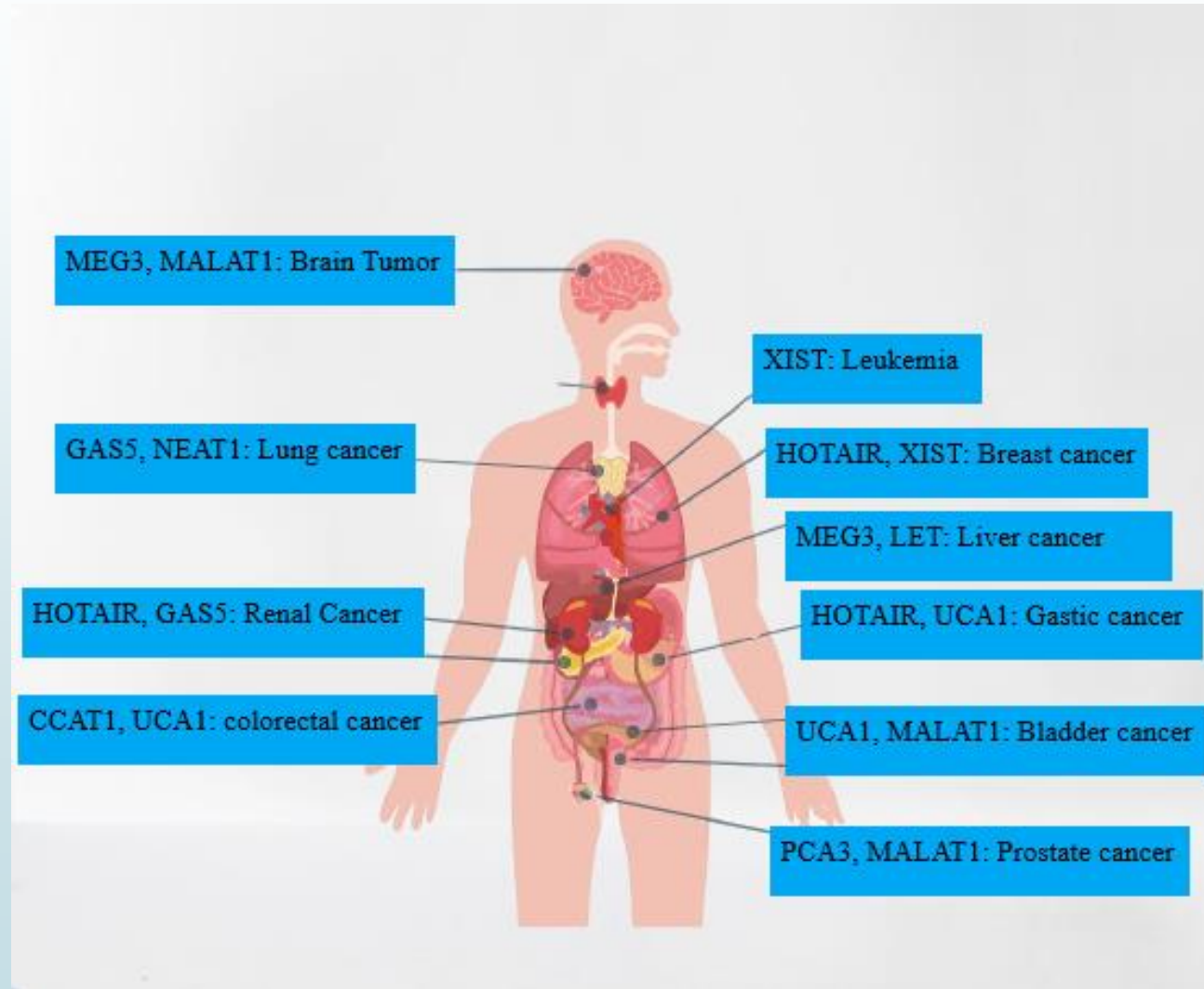
Università degli Studi di Palermo

Functions of Long non-coding RNAs



Taken from "State of The Art Technologies Used to Explore Long Non-coding RNAs in Cancer Disease", S. Salehi et al., Stat. J. Cell. Mol. Med., 20(10):1-21, 2017

LncRNAs have a role in human diseases

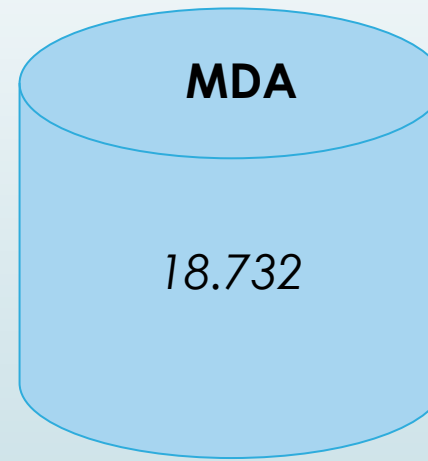


Problem

- Only a few experimentally verified lncRNA – disease associations are known



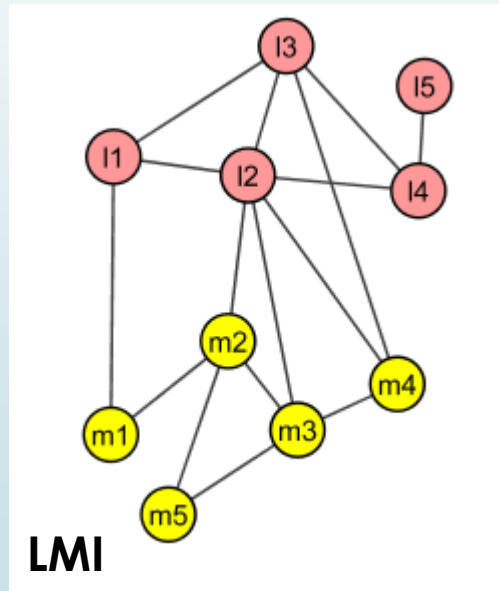
lncRNA – disease associations
stored in cuiLab db



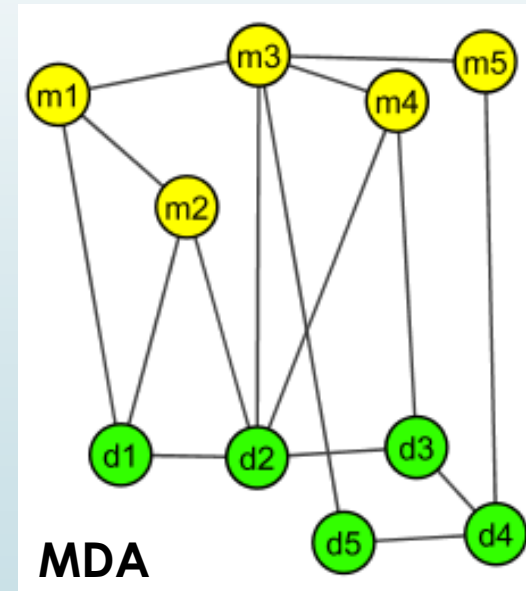
miRNA – disease associations
stored in *HMDD db*

Observation

- Large amounts of lncRNA-miRNA interactions and miRNA-disease associations have been collected in public databases



10.112 lncRNA-miRNA in
starBase db



18.732 miRNA-disease associations
In HMDD db



Computational approaches for lncRNA-disease associations prediction

Do not use known LDA

X.Chen., 2015, Scientific Reports

S. Alaimo et al. , 2014,
Frontiers in Bioengineering
and Biotechnology

Do use known LDA

C. Domeniconi et al. ,
2018, Bioinformatics

M.N.Wang et al. 2021,
Neurocomputing

.....

.....



Computational approaches for lncRNA-disease associations prediction

Do not use known LDA

X.Chen., 2015, Scientific Reports

S. Alaimo et al. , 2014,
Frontiers in Bioengineering
and Biotechnology

Do use known LDA

C. Domeniconi et al. ,
2018, Bioinformatics

M.N.Wang et al. 2021,
Neurocomputing

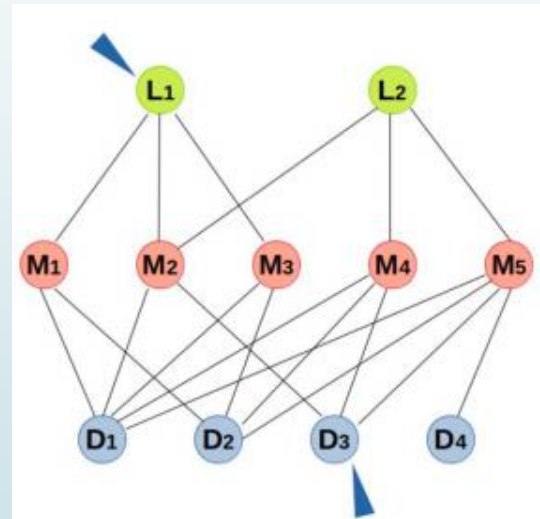
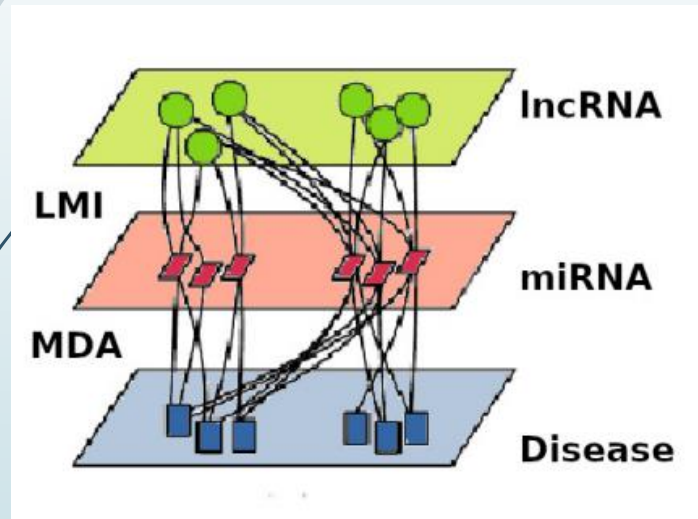
A dark blue arrow points to the right from the left edge of the slide. Below it, several thin, curved lines in shades of blue and grey sweep across the left side of the slide.

Proposed Approach

- ▶ Predict lncRNA-disease associations (LDAs) based on known lncRNA-miRNA interactions (LMIs) and miRNA-disease associations (MDAs)
- ▶ Consider miRNAs as intermediate molecules

Tripartite Graph Representation

Model LDAs, LMIs and MDAs as a tripartite graph



IncRNA 1.114

miRNA 1.065

diseases 885

The goal is to return a set of predicted LDAs

Neighborhood based approach

- For each pair lncRNA-disease (l,d) return in output an LDA with score:

$$S(l_i, d_j) = \alpha \cdot \frac{|M_{l_i} \cap M_{d_j}|}{|M_{l_i} \cup M_{d_j}|} + (1 - \alpha) \cdot \frac{|\bigcup_x (M_{l_x} \cap M_{d_j})|}{|\bigcup_x (M_{l_x}, M_{d_j})|}$$

- measures how much "connected" L and D are in the tripartite graph

M_{l_i} contains the number of miRNAs associated to l_i

M_{d_j} contains the number of miRNAs associated to d_j

Method

- ▶ Let T_{LMD} a tripartite graph
- ▶ We defined a prediction score LDA, based on neighborhood analysis with Apache Spark
- ▶ Statistical test based on recent experimental literatures : False Discovery Rate (FDR)



Datasets and gold standard

mirna-disease association (MDA 18.732) from HMDD db with:

- ▶ **1206 miRNAs and 894 diseases**

lncRNA-mirna interaction (LMI 10.112) from starBase db, with:

- ▶ 132 miRNAs and 1114 lncRNAs

lncRNA-disease association (LDA 1564) from cuilab db with:

- ▶ 914 lncRNA and 329 disease

A gold standard dataset from lncRNA disease db (X. Chen 2015) with:

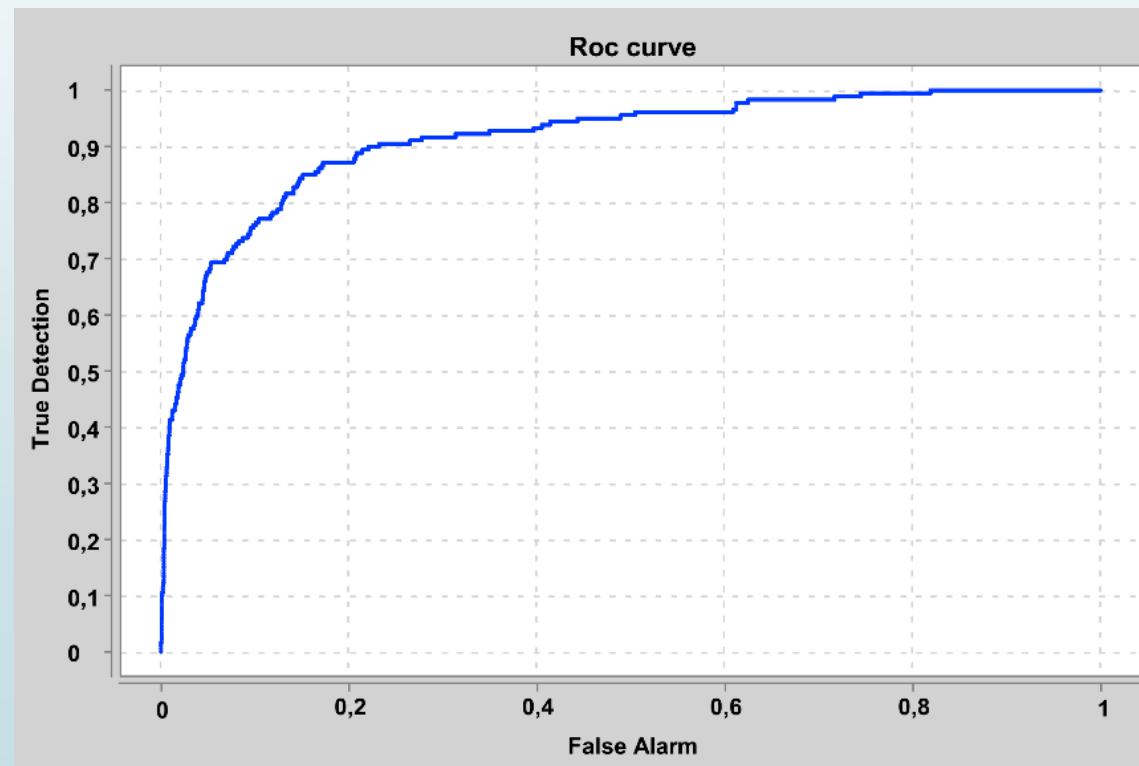
- ▶ 183 lncRNA disease

LOOCV:

- ▶ The predicted LDAs are ranked according to their *corrected* score
- ▶ Each verified LDA is left out in turn as test sample
- ▶ When the rank of this test sample exceeds a given threshold, the model provides a successful prediction
- ▶ At the varying of the threshold, compute true positive rate (TPR, sensitivity) and false positive rate (FPR, specificity)
 - ▶ Sensitivity: % of the test samples whose ranking is higher than the given threshold
 - ▶ Specificity: % of samples that are below the threshold
- ▶ We evaluated test by curve analysis ROC

ROC curve

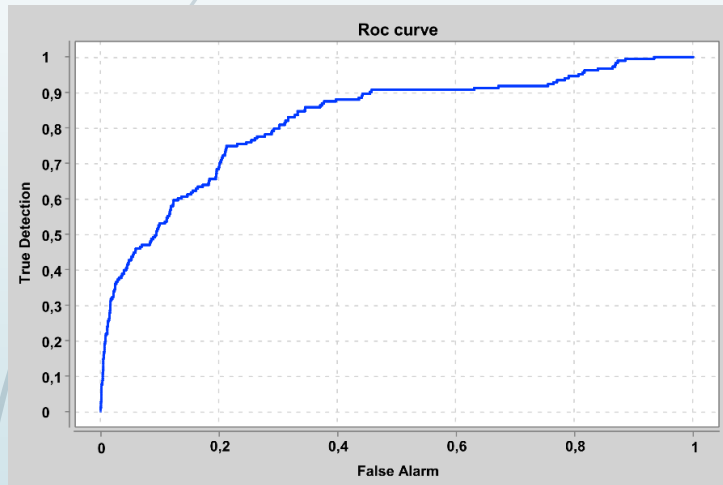
- ➔ Receiver-Operating Characteristics (ROC) curve is drawn by plotting TPR versus FPR at different thresholds



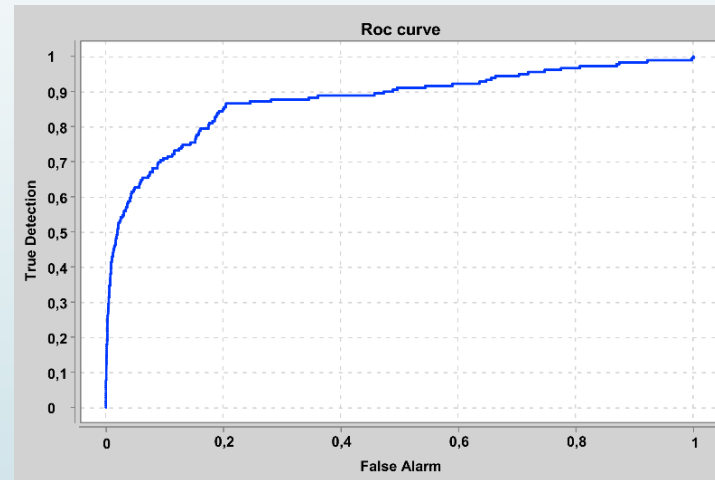
AUC = 0.91

Results with same datasets

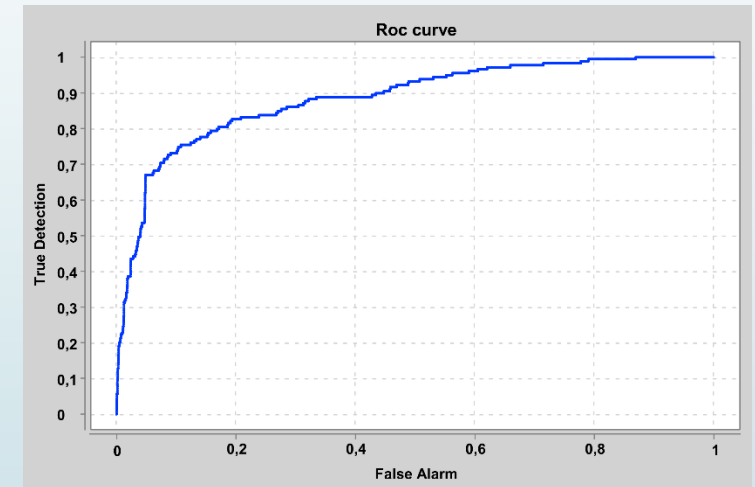
HMDD v3 dataset



Bonomo et al.
AUC = 0.91



Chen et al.
AUC = 0.87



Alaimo et al.
AUC = 0.88

Work in Progress

Do not use known LDA

X.Chen., 2015, Scientific Reports

S. Alaimo et al. , 2014, Frontiers in Bioengineering and Biotechnology



Do use known LDA

C. Domeniconi et al. , 2018, Bioinformatics

M.N.Wang et al. 2021, Neurocomputing

Work in Progress

Do not use known LDA

X.Chen., 2015, Scientific Reports

S. Alaimo et al. , 2014, Frontiers in Bioengineering and Biotechnology



Do use known LDA

C. Domeniconi et al. , 2018, Bioinformatics

M.N.Wang et al. 2021, Neurocomputing

.....





Related Publications

► Bonomo M.; La Placa A.; Rombo S.E. Prediction of lncRNA-Disease Associations from Tripartite Graphs. DOI:10.1007/978-3-030-71055-2_16. pp.205-210. In VLDB Workshops, Poly 2020 and DMAH 2020, Virtual Event, August 31 and September 4, 2020. In LECTURE NOTES IN COMPUTER SCIENCE - ISSN:1611-3349 vol. 12633

► Bonomo M.; La Placa A.; Rombo S.E. Prediction of Disease-lncRNA Associations via Machine Learning and Big Data Approaches. In Knowledge Modelling and Big Data Analytics in Healthcare, CRC Press, 2021.

Thank you!

