

Combining bottom up and top down approaches for Drug-Target-Interaction prediction

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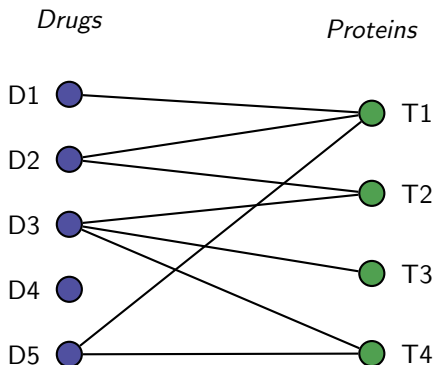
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Outline

1. Problem Description

Problem Description

Prediction over bipartite graph:



Classification of recent approaches¹²

	Drugs	Protein
bottom-up	<ul style="list-style-type: none">▶ GCN over molecules▶ drug similarity	<ul style="list-style-type: none">▶ secondary structure prediction▶ contact prediction▶ convolution over amino acid sequences
top-down	<ul style="list-style-type: none">▶ network approaches▶ drug similarity	<ul style="list-style-type: none">▶ protein similarity

¹Chen Wang et al., Briefings in Bioinformatics, 2018

²Yu Ding et al., Briefings in Bioinformatics, 2019

Problems of recent approaches

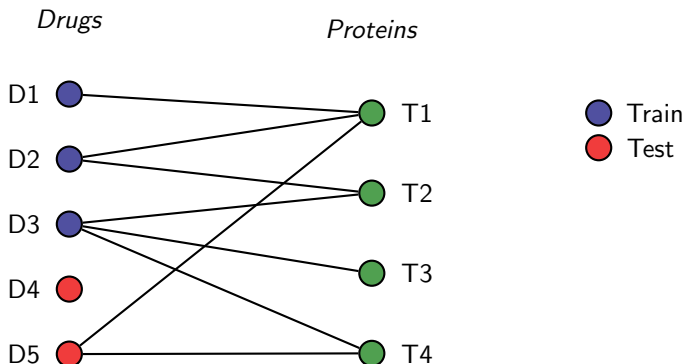
Main issues:

- ▶ Lack ability to generalize or are unable to spot small differences
- ▶ Usually only top-down or bottom-up
- ▶ Not making use of interaction networks

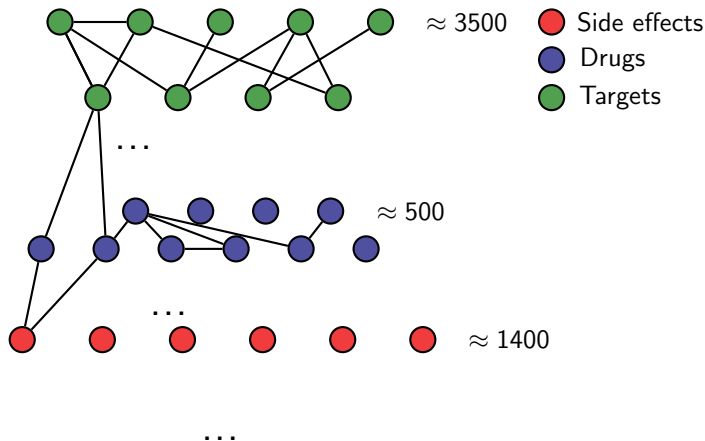
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Features in context



Available data structures

→ Combine Bottom up and top down features for drugs and features

Data structures used:

- ▶ PPIs, DTIs
- ▶ Motifs along proteins
 - ▶ find motif along protein sequences with HMMs for each drug
- ▶ DeepGO embeddings
- ▶ DDIs, SIDER, MedDRA
- ▶ DL2vec
- ▶ Seq2seq for drugs

Drugs

- ▶ `matrix[drug_index]`
- ▶ semantic similarities over MedDRA
- ▶ Seq2seq from Smiles
- ▶ DL2vec over SIDER-MedDRA-HPO

Proteins

- ▶ DeepGO
- ▶ Motifs from HMM
- ▶ DL2vec Embeddings from Azza and Jun
- ▶ node degree percentile (Gray encoding)
- ▶ GNNs (node features):
 - ▶ Motifs from HMM
 - ▶ DDI induced targets
 - ▶ semantic similarity weighted targets
 - ▶ DL2vec embeddings from Azza and Jun
 - ▶ DL2vec over HPO/MP/MedDRA
 - ▶ DL2vec over PhenomeNET (GO,MP/HPO, SIDER)

Architectures

- ▶ every possible layer from PyG
- ▶ Node features with/without GNN
- ▶ `cat([drug_feat, prot_feat])` as (node-) feature
- ▶ siamese over drugs and proteins/their (graph-)embeddings

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Not included:

- ▶ Variations of all datasets