Latent representations of chemical ligands to predict combinatorial receptor-ligand interactions

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Receptors process signals from the environment



Intracellular Response

Understanding combinatorial receptor-ligand interactions



Similar chemicals activate similar sets of receptors



Receptors

Hallem, Ho, Carlson. Cell 2004.

Motivation



Predicting ligand-receptor interactions

Ability to control intracellular behaviors

Challenge: representing inputs for predictive algorithms





Images

Text

Low-level representation schemes: molecular fingerprints



Low-level representation schemes: SMILES

Simplified Molecular-Input Line-Entry System



Machine learning models typically require high quantities of data



In the absence of a large dataset, feature abstraction is necessary

Recent trend: feature abstraction with variational autoencoders (deep neural network)



Current models

Grammar Variational Autoencoder

Kushner et al. 2017



Jin et al. 2018



Automatic Chemical Design Using a Data-Driven Continuous Representation of Molecules







Syntax-Directed Variational Autoencoder for Structured Data

Dai et al. 2018

Gómez-Bombarelli et al. 2016

Issues with existing SMILES-based models



Cc1cn2c(CN(C)C(=O)c3ccc(F)cc3C)c(C)nc2s1 Cc1cc(F)ccc1C(=O)N(C)Cc1c(C)nc2scc(C)n12

Current models overemphasize molecular geometry

GrammarVAE latent space visualization



Kushner et al. 2017

Evaluating the latent space of current models



Molecular fingerprint still performs the best

Data from Hallem, Ho, Carlson. Cell 2004.

Our Approach

Incorporating prior knowledge helps the model



Receptors

Our model: two-tower approach



Two-tower architecture



Our results





Combinatorial Ligand-receptor binding Feature abstraction with VAEs: Two-tower approach Predicting receptor activities