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Integrating *in silico* and *in vitro* data to identify putative thyrotropin- releasing hormone receptor ligands Mahmoud Shobair¹, Christopher Grulke¹, Daniel Chang¹, Ryan Lougee², Katie Paul Friedman¹, Ann Richard¹

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Abstract and Background

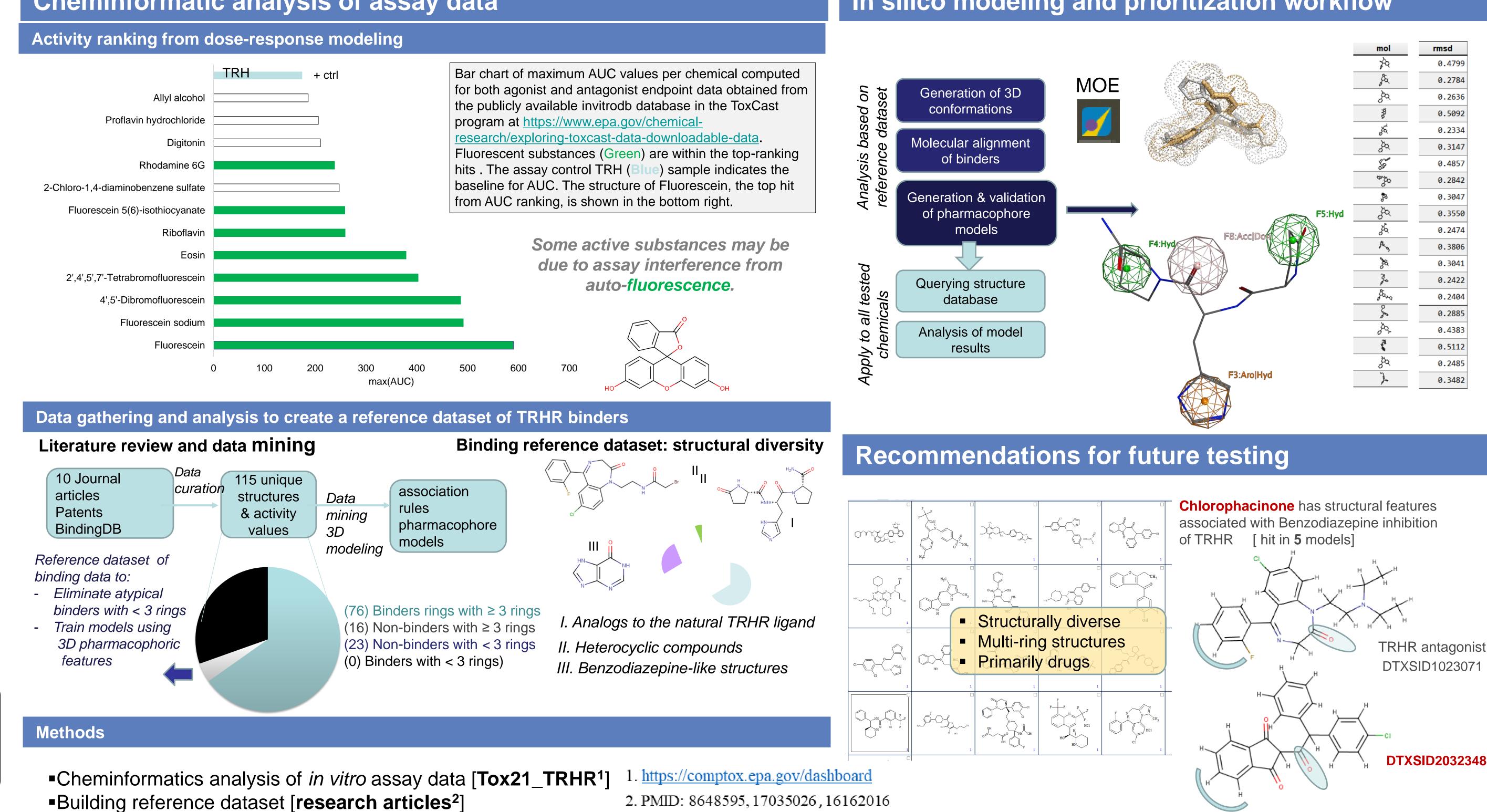
Despite progress in applying high-throughput screening (HTS) technologies to toxicology, exemplified by the Tox21 and ToxCast programs, the challenge of relating biochemical outputs to molecular initiating events (MIEs) and adverse outcome pathways (AOPs) remains challenging, particularly when the relationship between the biochemical output and MIE is indirect. Thyrotropin-Release Hormone Receptor (TRHR) activation is an MIE in the Thyroid-Hormone (TH) AOP, with unknown risk by environmentally-relevant chemicals. A Tox21 HTS biochemical TRHR assay is available but currently without orthogonal or confirmatory Tox21 or ToxCast assays to help confidently differentiate TRHR positive and negative responses. To determine if environmentally relevant chemicals can directly interact with TRHR, we developed an in-silico cheminformatic workflow to amplify biological signals and improve confidence and specificity in the TRHR assay results. The tiered approach: 1) identifies structure-activity patterns using chemotype-enrichment analysis; 2) filters noise from cytotoxicity or assay interference; and 3) prioritizes potential TRHR modulators by likelihood of binding. To build a training set, we created a curated reference dataset from literature studies of competitive binding that reported chemical concentration required to displace binding of radiolabeled TRH. The dataset is balanced between binders and non-binders and mainly contains TRH derivatives and psychoactive drugs. Using pharmacophore modeling, 3D descriptors discriminated between binders and non-binders. Preliminary results suggest that less than 11% of actives in the Tox21 assay contain TRH-like binding features, due to the latter reflecting conservation of the TRHR ligand binding site and structural similarity to known TRHR modulators, benzodiazepines and neuropeptides. The presented tiered workflow increases the value of *in vitro* data for chemical prioritization as it is grounded by the physical determinants directly related to relevant signals in experimental results. Our findings suggest that combining structure-based methods and data enrichment analysis can increase confidence in HTS results and define a scope of prediction for risk

Background: In vitro screening of Tox21 library for TRHR activity

1. Tox21_TRHR assay design TRH Fluorescence TRH Gq-PLC-IP3-IP3R → TRHR signal Large number of diverse TSHR is a GPCR with a few known 2. Assay hits environmental chemicals agonists or antagonists. screened, yet **false** and antagonist modes. negatives and positives Tox21_TRHR are expected. This assay measures agonism or (7872 total antagonism for TRHR through the Gq-Ca2+ Goal is to identify subset 160 antagonist 157 both chemicals adonist of likeliest true actives pathway. screened) from the full set of assay results 388 Total Hits The assay design includes potential sources 3. Prioritization of artefacts and non-specific interactions, as it indirectly measures TRHR activity leading to the sub-hypothesis: Approach is to prioritize subset of actives (true hits) and inactives (potential false negatives) for follow-up testing using: The screening assay may not be very specific for TRHR agonists, and antagonists. domain knowledge Objective: Create a workflow to identify & > chemotype enrichments prioritize chemicals with structural features associated with binding to TRH > in silico computational chemistry models binding site.

U.S. Environmental Protection Agency Office of Research and Development

Cheminformatic analysis of assay data



- •2D Filters/3D modeling [MOE³]
 - chemotype enrichments & heuristics

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In silico modeling and prioritization workflow

2. PMID: 8648595, 17035026, 16162016



MOE (The Molecular Operating Environment), software available from Chemical Computing Group Inc., 1010 Sherbrooke Street West, Suite 910, Montreal, Canada H3A 2R7 (http://www.chemcomp.com

For more information, contact: Mahmoud Shobair, PhD

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