

Workflow for Defining Reference Chemicals to Assess Performance of *In Vitro* Assays

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Abstract

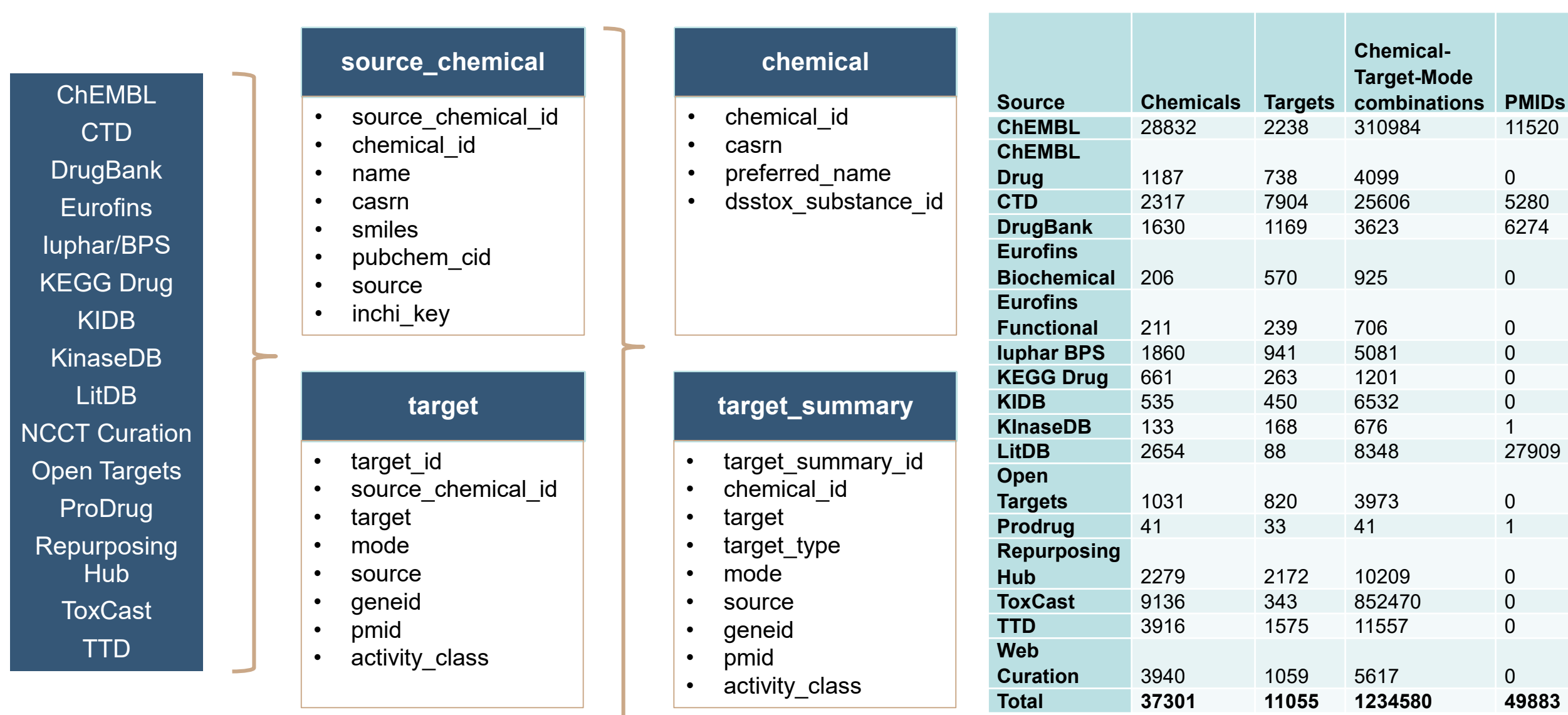
OBJECTIVE: Develop a semi-automated workflow for identifying and annotating reference chemicals to validate *in vitro* assay data.

BACKGROUND: Use of reference chemicals is key for validating high-throughput *in vitro* assay data used in predictive toxicology, but developing reference chemical lists has historically only occurred on a small scale due to context specificity and resource limitations.

METHODS: We extracted information from curated and non-curated open-source databases on molecular target, chemical, and mode of action into **RefChemDB**. We also compared data from EPA's ToxCast program to results from the literature. To determine support, we tallied independent reports of each unique chemical-target-mode combination. To contextualize support from different data sources, we manually validated a subset of the data that is associated with PubMed IDs and determine precision rates across hand curators.

RESULTS: We compiled a database with 1,234,580 unique chemical-target-mode combinations. Performance of ToxCast bioassays strongly correlated with the level of support for a chemical-target-mode combination. We hand-curated data for 54 molecular targets, with precision rates of 82.7% from curated sources and 39.5% from automated literature extraction, informing our workflow recommendation. A list of candidate reference chemicals was created by selecting chemical-target-mode combinations with a minimum support level of 5 records.

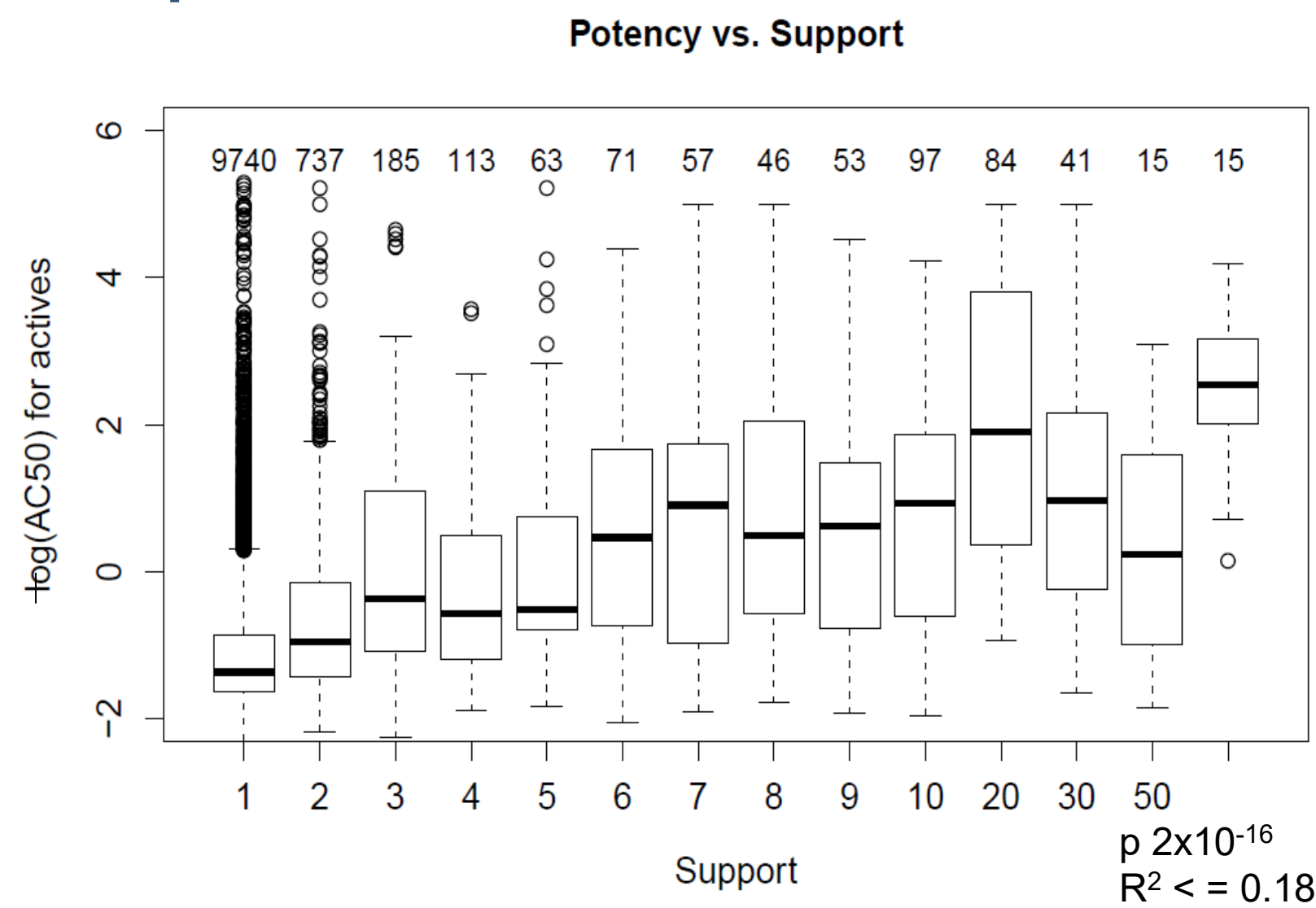
Database Structure



RefChemDB structure and contents.

Results

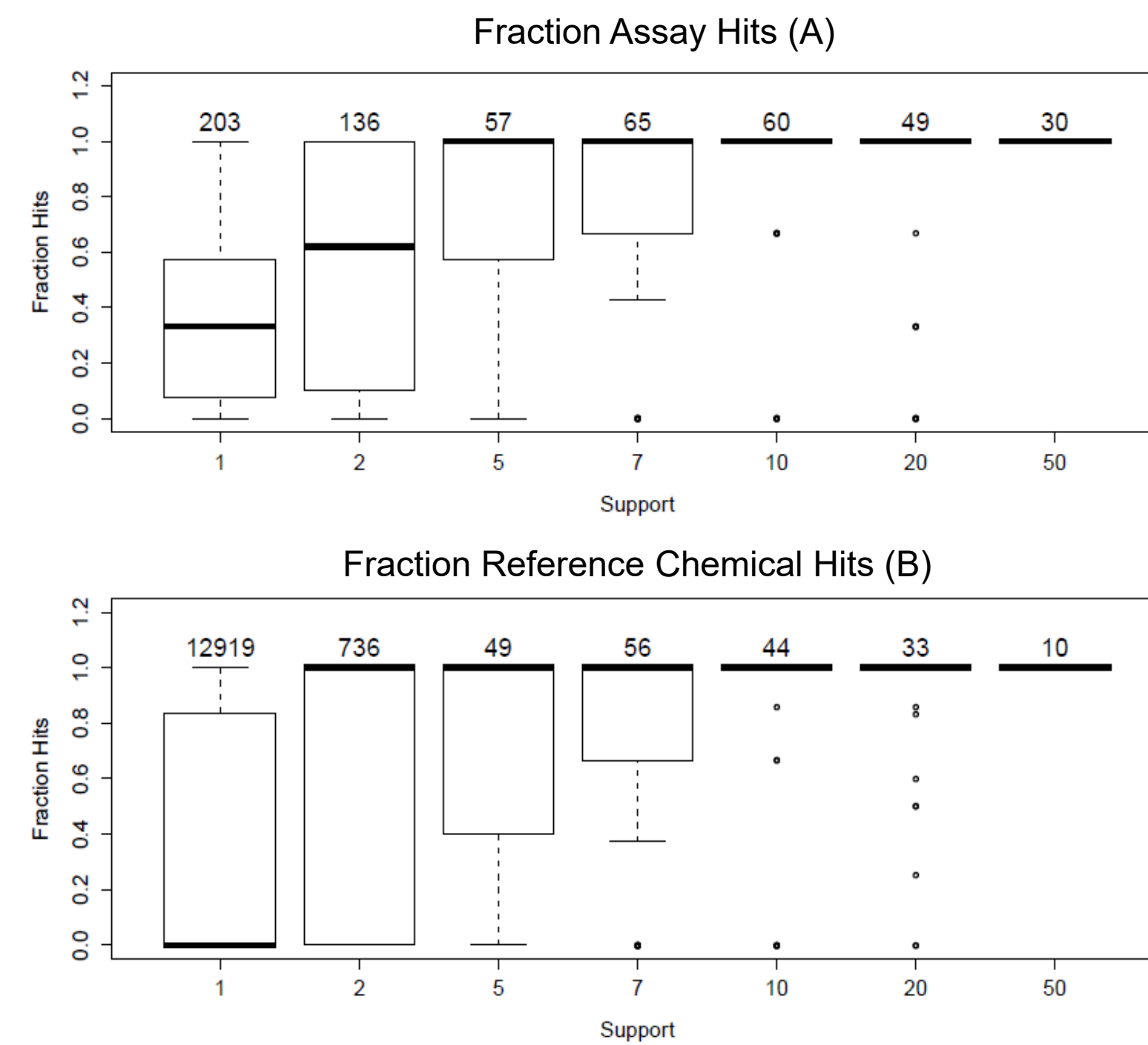
Comparison with ToxCast



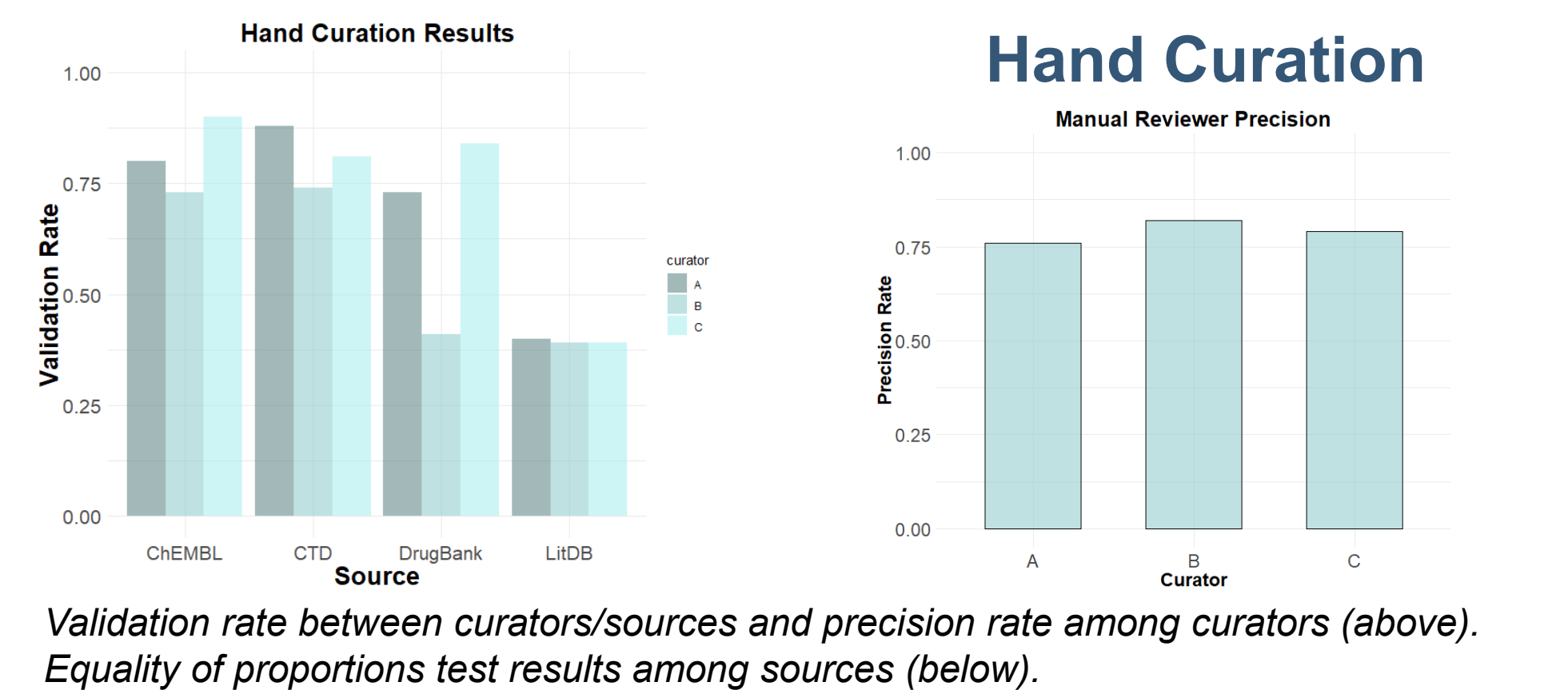
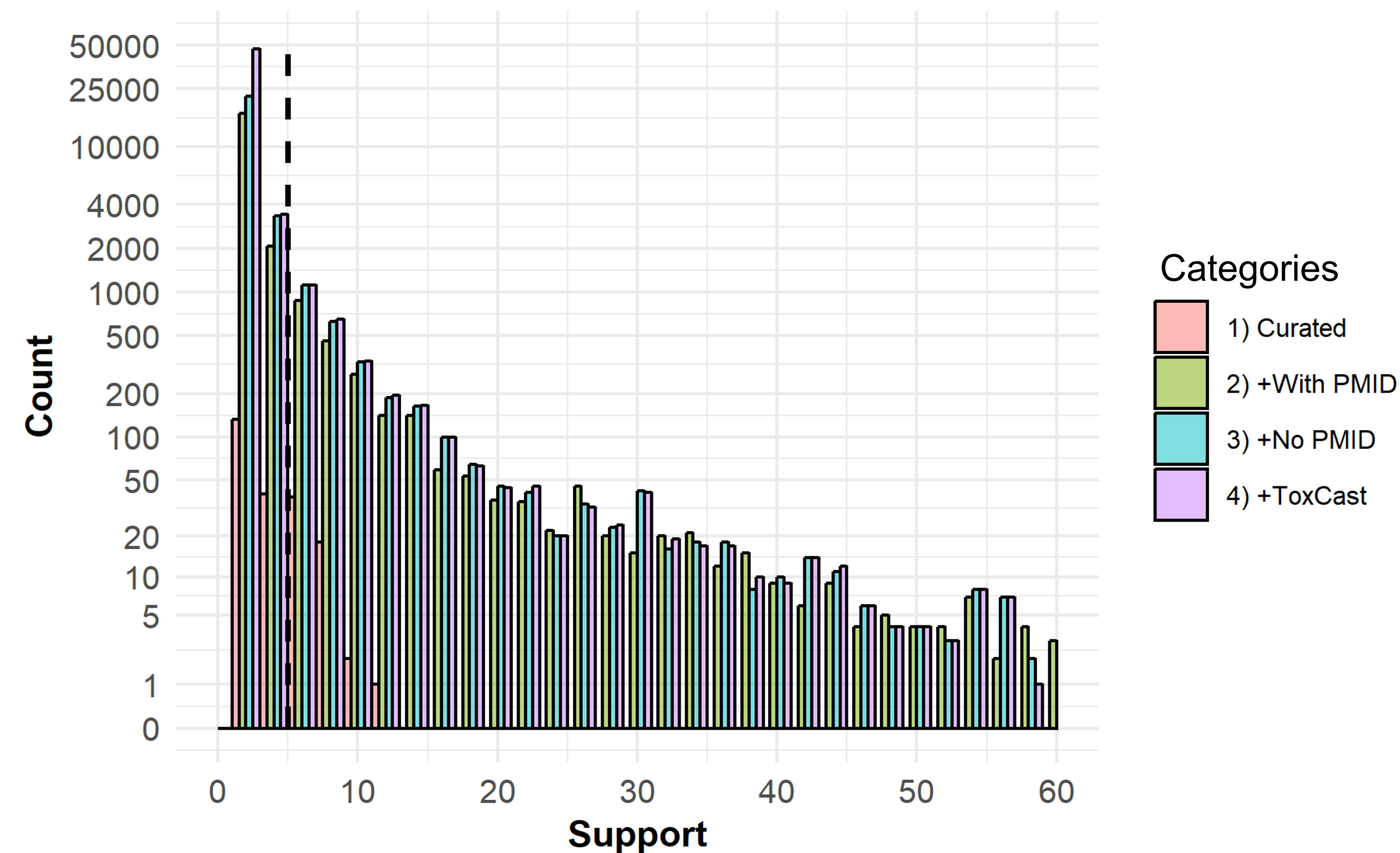
Candidate Reference Chemicals

Target Symbol	Target Name	Chemicals	Support
CA2	Carbonic Anhydrase 2	106	2453
CA1	Carbonic Anhydrase 1	105	1981
ESR1	Estrogen Receptor 1	85	1371
DRD2	Dopamine Receptor D2	81	952
AR	Androgen Receptor	63	750
ESR2	Estrogen Receptor 2	41	664
PTGS2	Prostaglandin-Endoperoxide Synthase 2	46	646
OPRM1	Opioid Receptor Mu 1	48	638
CA9	Carbonic Anhydrase 9	23	512
PPARA	Peroxisome Proliferator-Activated Receptor Alpha	27	511
PPARG	Peroxisome Proliferator-Activated Receptor Gamma	26	486
NR3C1	Nuclear Receptor Subfamily 3 Group C Member 1	37	482
HTR2A	5-Hydroxytryptamine Receptor 2A (Serotonin Receptor 2A)	43	476
ADRB2	Beta-2 Adrenergic Receptor	44	472
ACHE	Acetylcholinesterase	28	470
SLC6A4	Solute Carrier Family 6 Member 4	40	423
ABCB1	ATP Binding Cassette Subfamily B Member 1	35	418
KCNH2	Potassium Voltage-Gated Channel Subfamily H Member 2	48	412
HRH1	Histamine Receptor H1	45	399
HDAC1	Histone Deacetylase 1	12	387

Top 20 most-supported targets with tally of chemical interactions observed in literature with total support.



Support Across 4 Categories



Source 1	Source 2	p-value
ChEMBL	CTD	0.574
ChEMBL	DrugBank	0.0337*
CTD	DrugBank	0.7761
ChEMBL	LiDB	<2.2e-16***
CTD	LiDB	<2.2e-16***
DrugBank	LiDB	<2.2e-16***

Summary

- Query RefChemDB**
 - Return target, chemical, mode, and source information from the database
- Evaluate reference candidate lists**
 - Identify chemicals with support > 5
 - Identify chemicals in highly accurate sources (DrugBank, ChEMBL)
- Validate data**
 - Use PubMed IDs where possible to confirm assay data and metadata with two reviewers
 - Finalize with subject matter expert review
- Deep dive into literature**
 - Use literature search tools like EPA's PubMed Abstract Sifter to find more definitive target/chemical interaction information

To access RefChemDB files, please see Judson et al. 2018 in ALTEX

Resources

[1] Jaworska, J., N. Nikolova-Jeliazkova and T. Aldenberg (2005). "QSAR applicability domain estimation by projection of the training set descriptor space: a review." *Altern Lab Anim* **33**(5): 445-459. [2] NICEATM / ICCVAM. (2011). "NICEATM DRAFT ED BRD: BG1Luc ER TA Test Method – Section 3.0." [3] Mansouri, K., A. Abdelaziz, A. Rybacka, A. et al. (2016). "CERAPP: Collaborative Estrogen Receptor Activity Prediction Project." *Environ Health Perspect* **124**(7): 1023-1033. [4] Tiikkainen, P., L. Bellis, Y. Light and L. Franke (2013). "Estimating error rates in bioactivity databases." *J Chem Inf Model* **53**(10): 2499-2505. [5] Casey, Chang, et al. (2018). "Evaluation and Optimization of Pharmacokinetic Models for *In Vitro* to *In Vivo* Extrapolation of Estrogenic Activity for Environmental Chemicals." *Environ. Health Perspect* **126**(9): 097001-14. [6] Baker, N., Knudsen, T., Williams, A. (2017). "Abstract Sifter: a comprehensive front-end system to PubMed." *F1000Res* **6**: 2164.