

Molecular Point of Departure (mPOD) Determination From *In Vitro* High-Throughput Transcriptomics Data.

Joshua A. Harrill

USEPA Center for Computational Toxicology and Exposure (CCTE)





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NAMs-Based Tiered Hazard Evaluation Approach

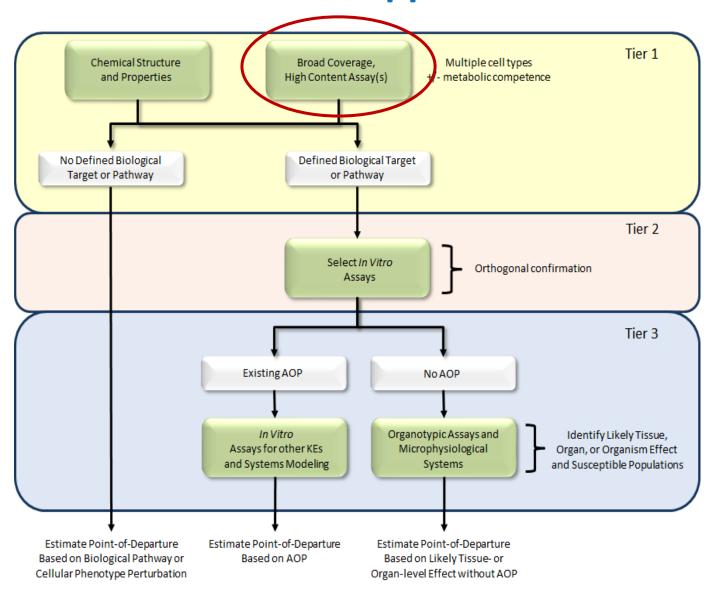
High throughput profiling (HTP) assays are proposed as the first tier in a NAMs-based hazard evaluation approach.

HTP Assay Criteria:

- 1. Yield bioactivity profiles that can be used for potency estimation, mechanistic prediction and evaluation of chemical similarity.
- 2. Compatible with multiple human-derived culture models.
- 3. Concentration-response screening mode.
- 4. Cost-effective.

To date, EPA has identified and implemented two HTP assays that meet this criteria.

- High-Throughput Transcriptomics [HTTr]
- High-Throughput Phenotypic Profiling [HTPP]



The NexGen Blueprint of CompTox at US EPA

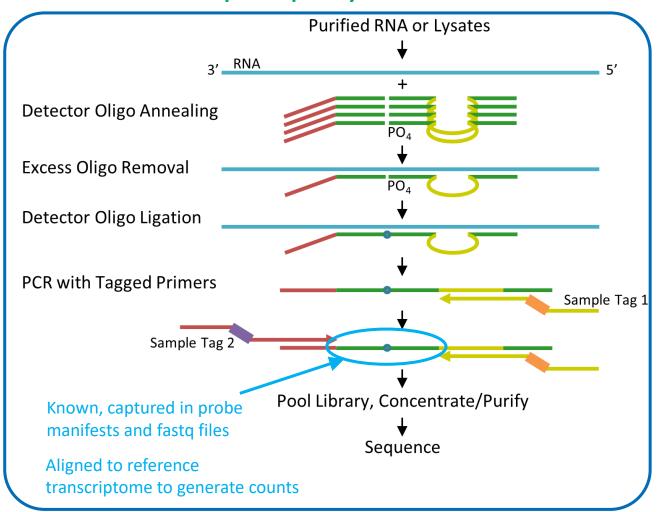
Thomas et al. (2019) DOI: 10.1093/toxsci/kfz058



Templated Oligo with Sequencing Readout (TempO-Seq)

- The TempO-Seq human whole transcriptome assay measures the expression of greater than 20,000 transcripts.
- Requires only picogram amounts of total RNA per sample.
- Compatible with purified RNA samples or cell lysates.
- Lysates are barcoded according to sample identity and combined in a single library for sequencing using industry standard instruments.
- Scalable, targeted assay:
 - 1) specifically measures transcripts of interest
 - 2) ~50-bp reads for all targeted genes
 - 3) requires less flow cell capacity than RNA-Seq

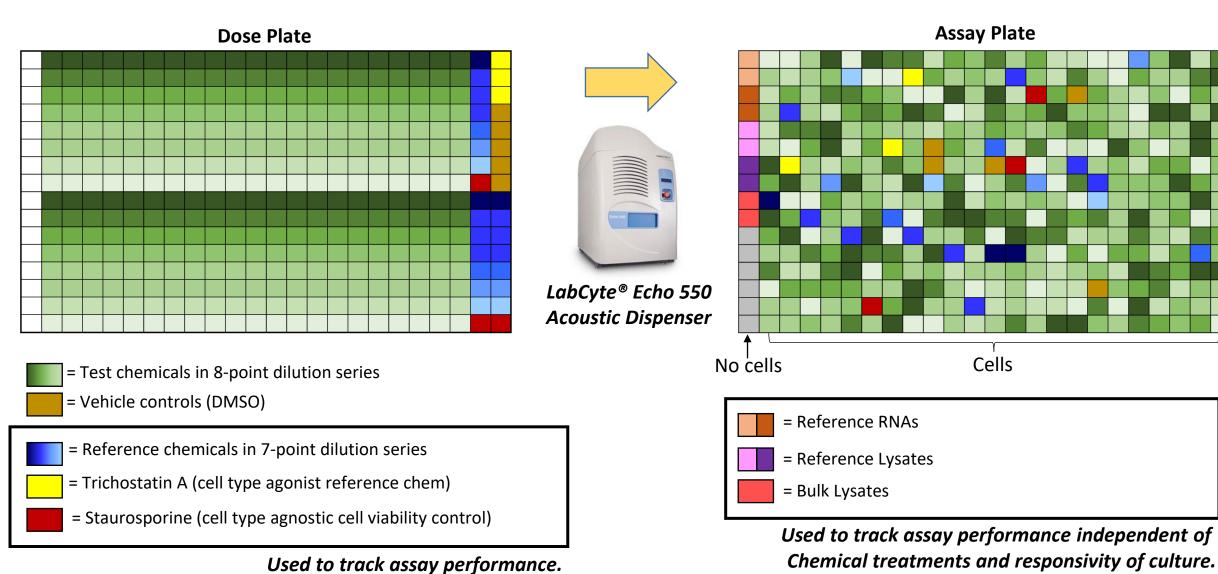
TempO-Seq Assay Illustration



Yeakley et al. (2017) DOI: 10.1371/journal.pone.0178302



Generic Experimental Design for HTTr



= Reserved for sequencing vendor



MCF7 Pilot Experimental Design

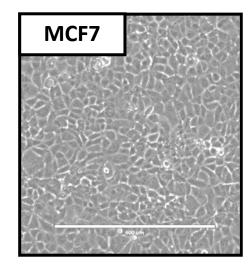
High-Throughput Transcriptomics Platform for Screening Environmental Chemicals

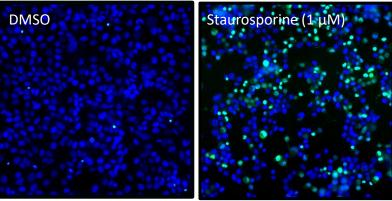
Joshua A. Harrill ,**,1* Logan J. Everett,* Derik E. Haggard ,**,1* Thomas Sheffield,**,1* Joseph L. Bundy,* Clinton M. Willis,**,2* Russell S. Thomas ,* Imran Shah ,* and Richard S. Judson .**

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doi: 10.1093/toxsci/kfab009 Advance Access Publication Date: 4 February 2021 Research Article

Parameter	Multiplier	Notes
Cell Type(s)	1	MCF7
Assay Formats:	2	High-Throughput Transcriptomics Cell Viability
Culture Condition	1	DMEM + 10% HI-FBS
Chemicals	44	ToxCast chemicals
Time Points:	1	6 hours
Concentrations:	8	3.5 log ₁₀ units; semi log ₁₀ spacing
Biological Replicates:	3	Independent cultures





CellEvent Caspase 3/7



MCF7 Pilot Chemical List

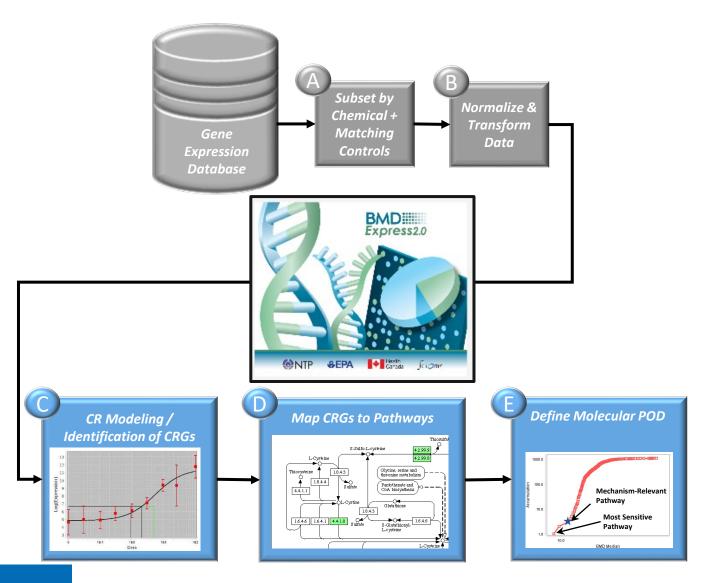
Table 1. Chemicals Used i	n the Study		
Name	Target Annotation	Name	Target Annotation
Cyproterone acetate Flutamide Nilutamide Vinclozolin Amiodarone hydrochlorid Cladribine 4-Cumylphenol 4-Nonylphenol, branched Bisphenol A Bisphenol B 4-Hydroxytamoxifen Clomiphene citrate (1:1) Fulvestrant Cyproconazole Imazalil Prochloraz Propiconazole Atrazine Cyanazine Simazine Buta fena cil Fomesafen Lactofen	DNA synthesis inhibitor ER agonist	Lovastatin Simvastatin Maneb Thiram Ziram Reserpine Rotenone Pyraclostrobin Trifloxystrobin Fenpyroximate (Z, E) Clofibrate Fenofibrate Farglitazar Perfluorooctanoic acid (PFOA) Perfluorooctanesulfonic acid (PFOS) Troglitazone Cycloheximide Bifenthrin Cypermethrin Tetrac 3,5,3'-triiodothyronine	HMGCR inhibitor Inhibition of metal-dependent and sulfhydryl enzyme systems Inhibition of metal-dependent and sulfhydryl enzyme systems Inhibition of metal-dependent and sulfhydryl enzyme systems Inhibition of the ATP/Mg2+ pump Mitochondria (complex I inhibitor) Mitochondria (complex III inhibitor) Mitochondria (complex III inhibitor) Mitochondrial electron transport inhibitor PPARa agonist, upregulates extrahepatic lipoprotein lipase PPARa agonist, upregulates extrahepatic lipoprotein lipase PPARa, pPARa agonist Protein synthesis inhibitor Sodium channel modulator T4 synthesis inhibitor THR agonist

- Chemicals were selected that cover a broad range of molecular targets with some redundancy within target class.
- Intentionally selected some chemicals whose molecular targets are not expressed in MCF7 cells (or in mammalian tissues).

Harrill et al. (2021) DOI: <u>10.1093/toxsci/kfab009</u>



BMDExpress for mPOD Determination



Based on National Toxicology Program Approach to Genomic Dose-Response Modeling (NTP RR 5)

BMDExpress Parameter	Criteria	
Pre-filter:	FC > 2 at any test concentration	
Models	Hill, Power, Linear, Poly2, Exponential 2 3 4 5	
BMR Factor:	1.349*SD of controls (10%)	
Best Model Selection:	Lowest AIC	
Hill Model Flagging:	'k' < 1/3 Lowest Positive Dose Exclude Flagged Hill from Best Model Selection	
Conc-Response Hit Criteria	(0.1*lowest conc. < BMC < highest conc.) BMC fit p-value > 0.1 BMCL / BMCU < 40	
Gene Set Analysis:	2 3 Concentration-responsive genes5% Gene Set Coverage	
Gene Set Collections:	MSigDB (Liberzon et al. 2015) BioPlanet (Huang et al. 2019) CMAP (Subramanian et al. 2005)	
Molecular Point of Departure	Most Sensitive Gene Set	

Harrill et al. (2021) DOI: <u>10.1016/j.cotox.2019.05.004</u>



Modeling of Signature Scores for mPOD Determination (1)

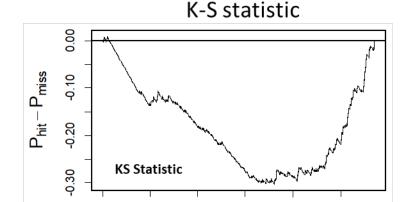
Step 1: Inputs

Experimental Data: Chemical_Conc \times Gene matrix of log_2 (fold-change) (l2fc) values.

Signature Collections: MSigDB (Liberzon et al. 2015), BioPlanet (Huang et al. 2019), CMAP (Subramanian et al. 2005)

Step 2: Pathway Scoring

Scores based on single sample GSEA method (Barbie et al. 2009)



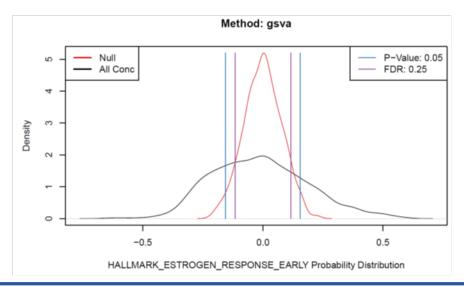
2000

Chemical_Conc × Pathway matrix of scores.

Gene Rank

Step 3: Cut-off Estimation via NULL Modeling

- For each gene, resample 12fc based on the crosssample gene distribution
- Calculate pathway scores for "null" data
 - One null distribution (n = 1000 scores) / pathway

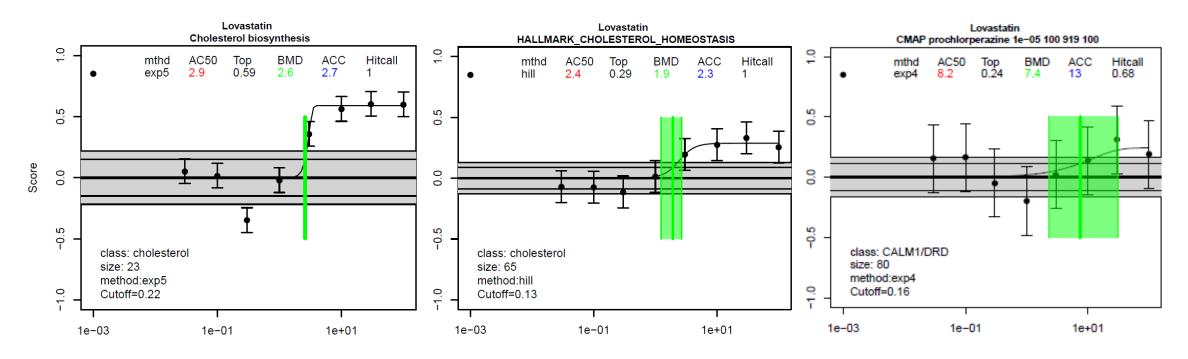




Modeling of Signature Scores for mPOD Determination (2)

Step 4: CR Modeling

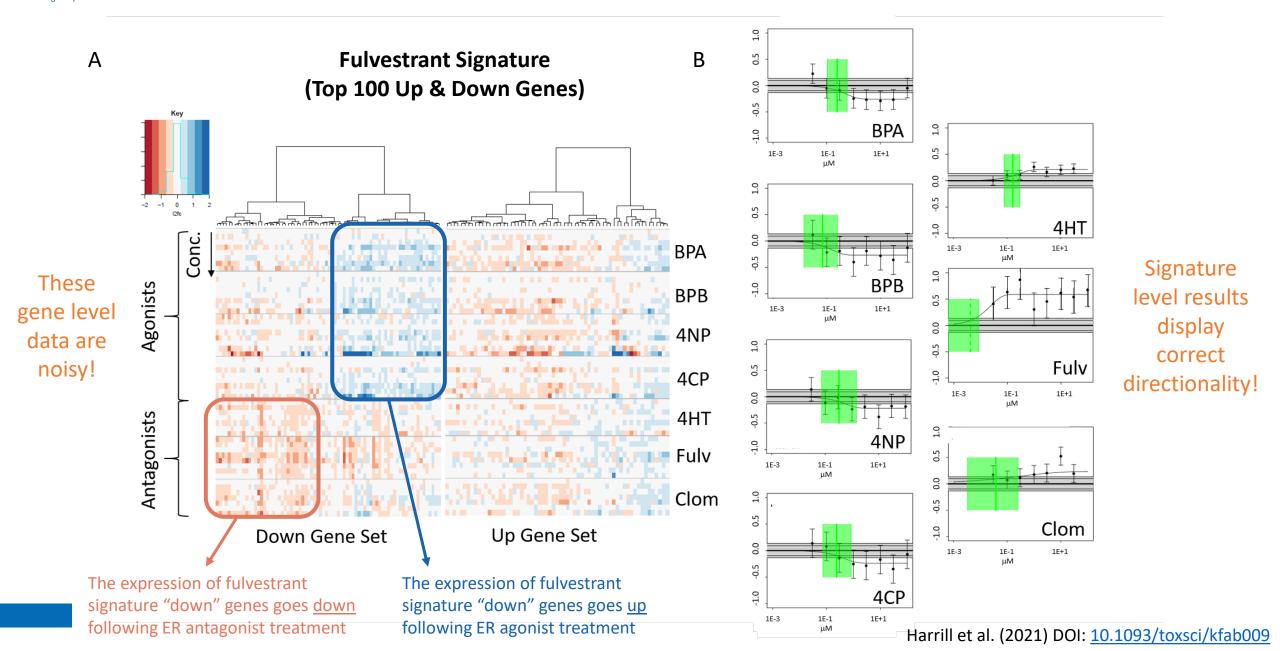
Concentration response modeling of signature scores using tcplfit2 (Sheffield et al. (2021) 10.1093/bioinformatics/btab779)



- Takes into account coordinated changes in gene expression that may not be identified using gene level fitting approaches.
- All curve forms from BMDExpress, plus constant model.
- Provides continuous hit calls for identifying high confidence and low confidence hits.



MCF7 Pilot Results: Directionality of Signature Scores





MCF7 Pilot Results: Comparison of mPOD Approaches

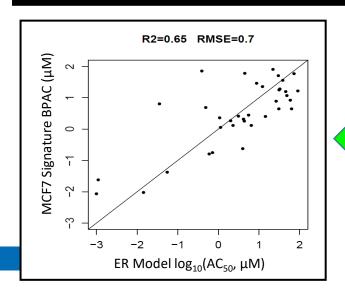


BPAC_{BMDX} → Most sensitive signature / pathway

BPAC_{HTS} → Lower 5th percentile of active AC50 values for ToxCast assays that pass a series of quality filters.

 $BPAC_{HTS}$ and $BPAC_{Sig}$ are in better agreement than $BPAC_{HTS}$ and $BPAC_{BMDX}$

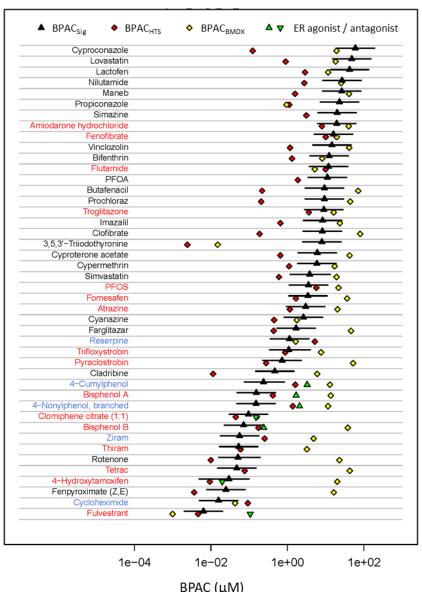
In most of these cases, $BPAC_{HTS}$ is also more potent than $BPAC_{BMDX}$.



Signature-based BPACs in MCF7 are concordant with ToxCast estrogen receptor (ER) model predictions.

 Brown et al. (2015) DOI: 10.1021/acs.est.5b02641

BPAC = Biological Pathway Altering Concentration



Harrill et al. (2021) DOI: 10.1093/toxsci/kfab009



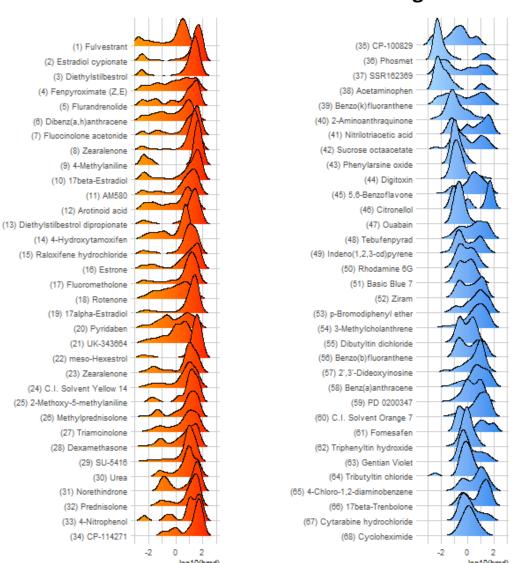
MCF7 HTTr Screening Results (1)

Median BMC of Active Signatures ($\log_{10} \, [\mu M])$

1784 Chemicals Screened # of Active Signatures 250 1000 _38 10 1001000 # of Active Signatures

5th % BMC of Active Signatures (log₁₀ [μM])

Distribution of BMCs of Active Signatures

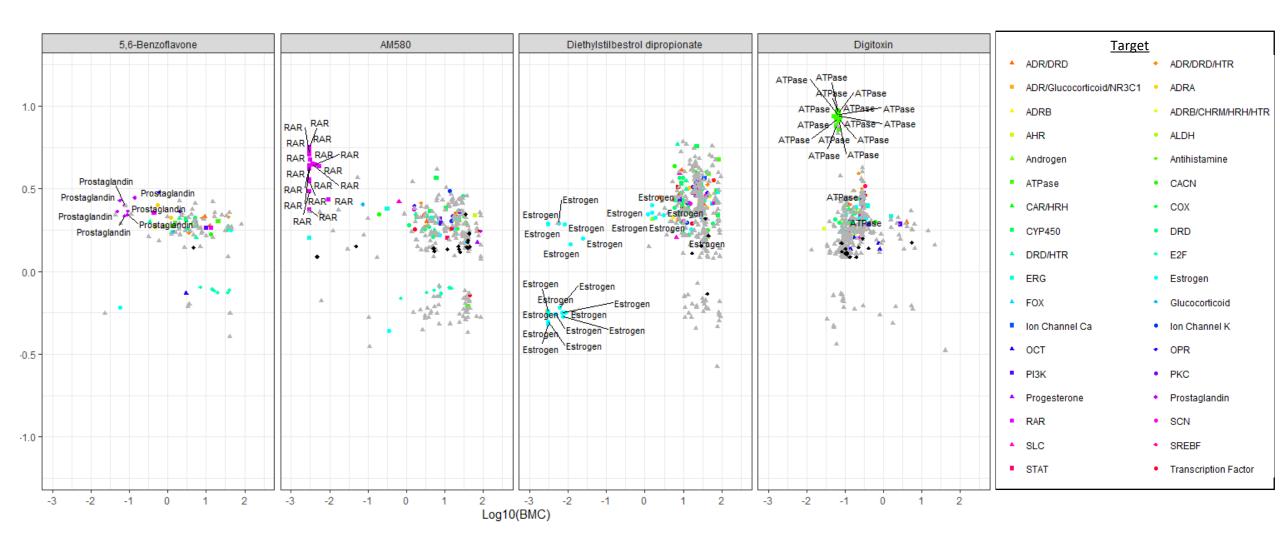


Other potent toxicants organometallics, dyes, etc) cause many signatures to be affected near the onset of biological activity.

Chemicals with known pharmacological targets show an "early wave" of biological activity.



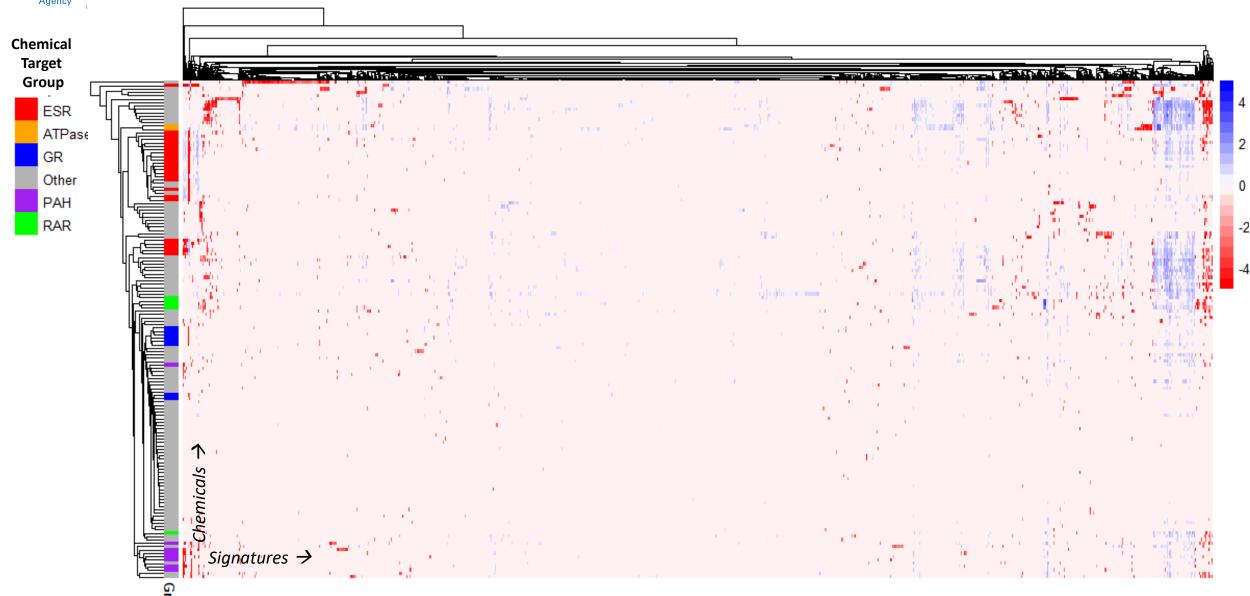
MCF7 HTTr Screening Results (2)



The most potent and efficacious signature hits correspond to known mechanisms for these chemicals.

United States Environmental Protection Agency

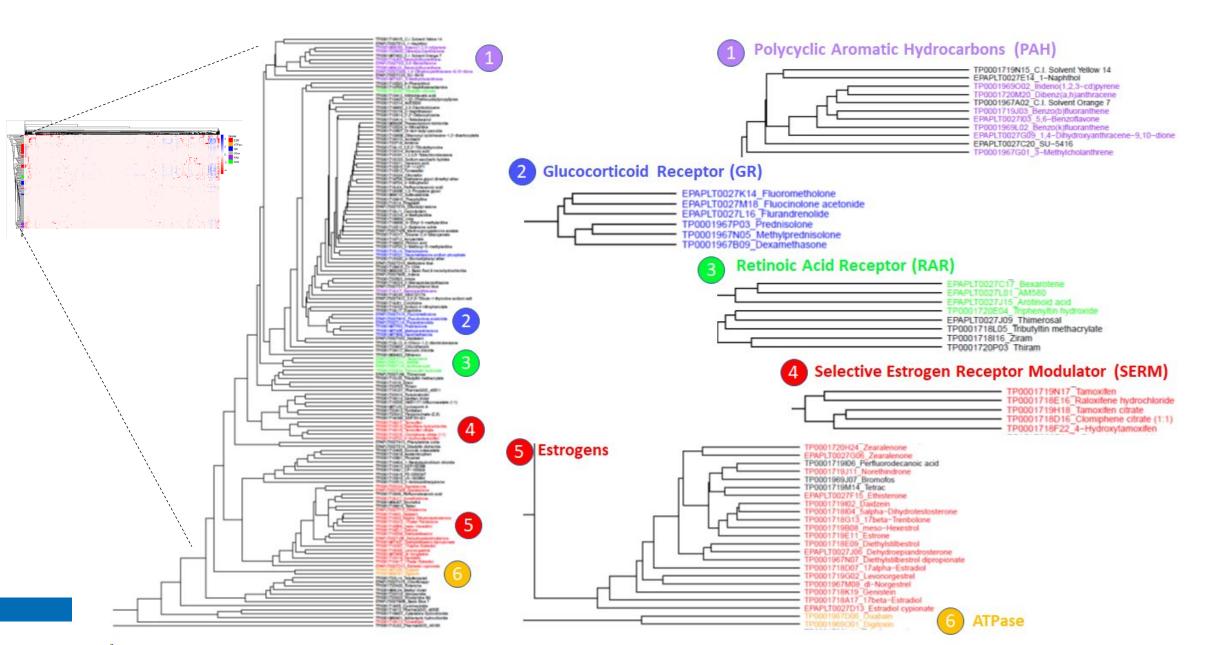
MCF7 HTTr Screening Results (3)



Clustering based on signed area under the curve (AUC) groups similar chemicals together.

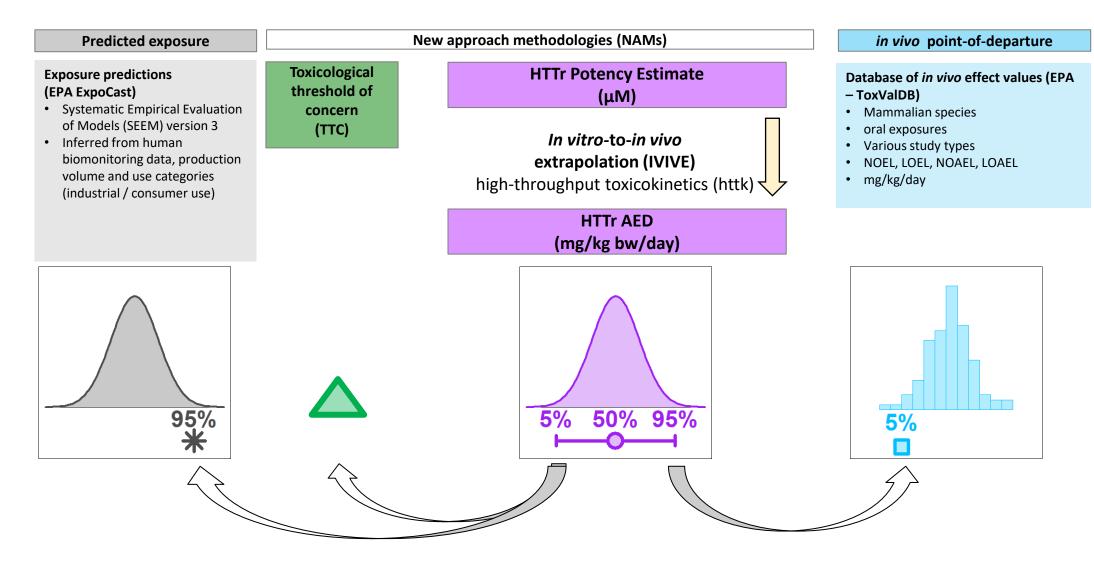


MCF7 HTTr Screening Results (4)





In Vitro to In Vivo Extrapolation (IVIVE) Using High-Throughput Toxicokinetic (httk) Modeling

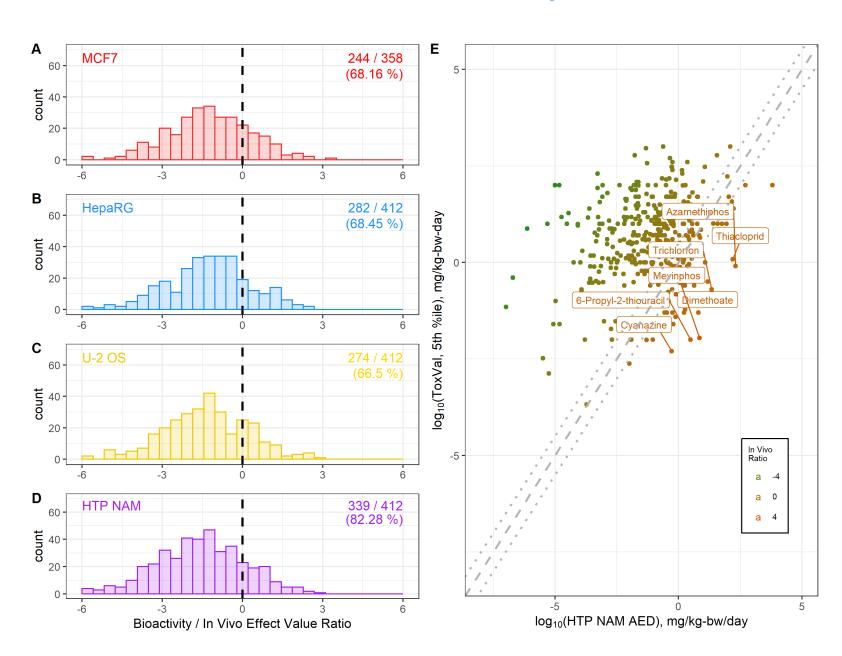


POD: point-of-departure
AED: administered equivalent dose



Bioactivity / In Vivo Effect Value Ratio Analysis

- Negative ratios indicate that AEDs derived from HTP NAMs molecular PODs are conservative surrogates for traditional in vivo PODs.
- When cell lines are considered individually,
 66-68% of chemicals had negative ratios.
- When considered in combination, the number and percentage of chemicals with negative ratios increased (82.3 %).
- Paul Friedman et al. (2020) (PMID: <u>31532525</u>)
 - Using ToxCast, 89 % of APCRA chemicals had negative ratios.
- When multiple cell types are considered, mPODs from HTTr screening appear to be conservative surrogates for *in vivo* PODs.
- Correlation of in vitro and in vivo is low.





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