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### **INTRODUCTION**

- 700+ commercial chemicals are introduced to the US market each year. Over 85% are approved for manufacturing despite the lack of experimental health and safety data.
- EPA ToxCast initiative, launched in 2007, uses high-throughput screening assays to expose living cells to chemicals to evaluate potential toxic and adverse health effects.
  - Phase I & II included 1859 chemicals
- There is a need for computational approaches to utilize existing toxicological data to then predict thresholds of untested chemicals and identify assays for assessing toxicity: estrogen agonist activation and responses to oxidative stress such as:
  - Chemical toxicity distributions (CTDs) utilized to perform probabilistic hazard assessments (PHAs)

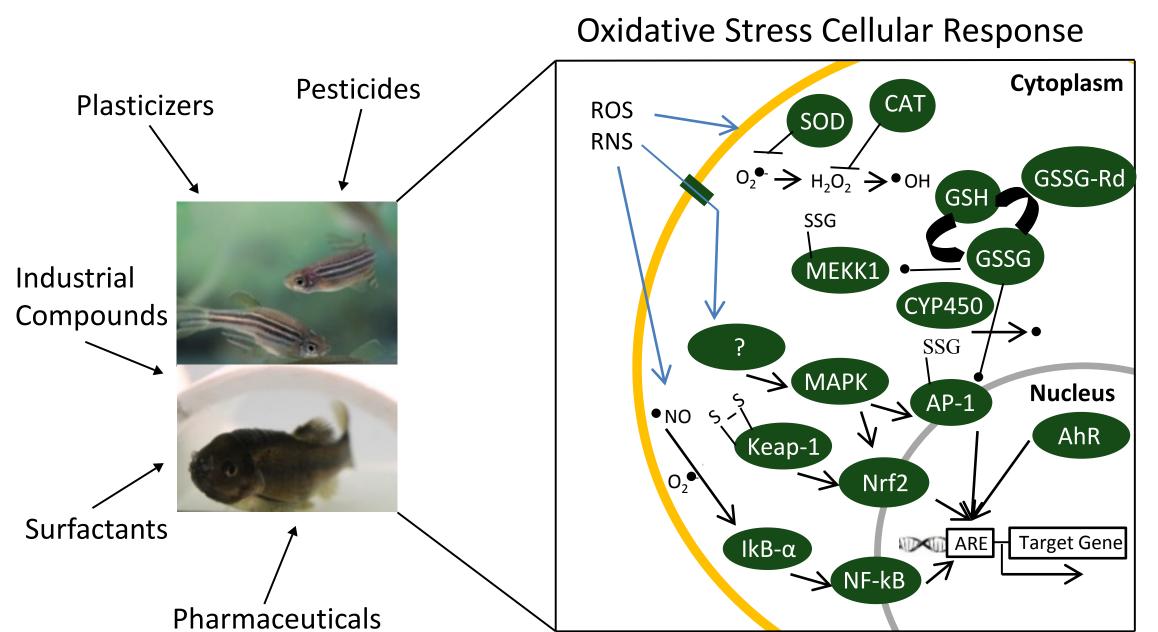


Figure 1: General oxidative stress mechanism. Developed from J. Limón-Pacheco and M.E. Gonsebatt Mutation Research 2009; 674: 137-147. Photo credit: Samuel P. Haddad.

## **STUDY OBJECTIVES**

- Using CTDs, develop PHAs for chemicals tested by ToxCast and the Endocrine Disruptor Screening Program (EDSP)
- Identify compounds with the highest potency to activate the estrogen receptor and elicit oxidative stress
- Examine the comparative sensitivity of *in vitro* assays for activity representing a diverse group of chemicals

# **MATERIALS AND METHODS**

- Data were compiled from the U.S. EPA's ToxCast database for *in vitro* estrogen agonist and oxidative stress related assays
- Data from three estrogen agonist and five oxidative stress response *in vitro* promoter assays are presented; descriptions are available in Table 2.
  - Utilized human liver and kidney cells to measure responses
- Percent Rank was assigned to each compound using a Weibull formula:

$$j = (i * 100) / (n + 1)$$

• CTDs for each assay were constructed by plotting  $EC_{50}$  values on a logarithmic scale versus a probability scale following methods previously described [1,2]

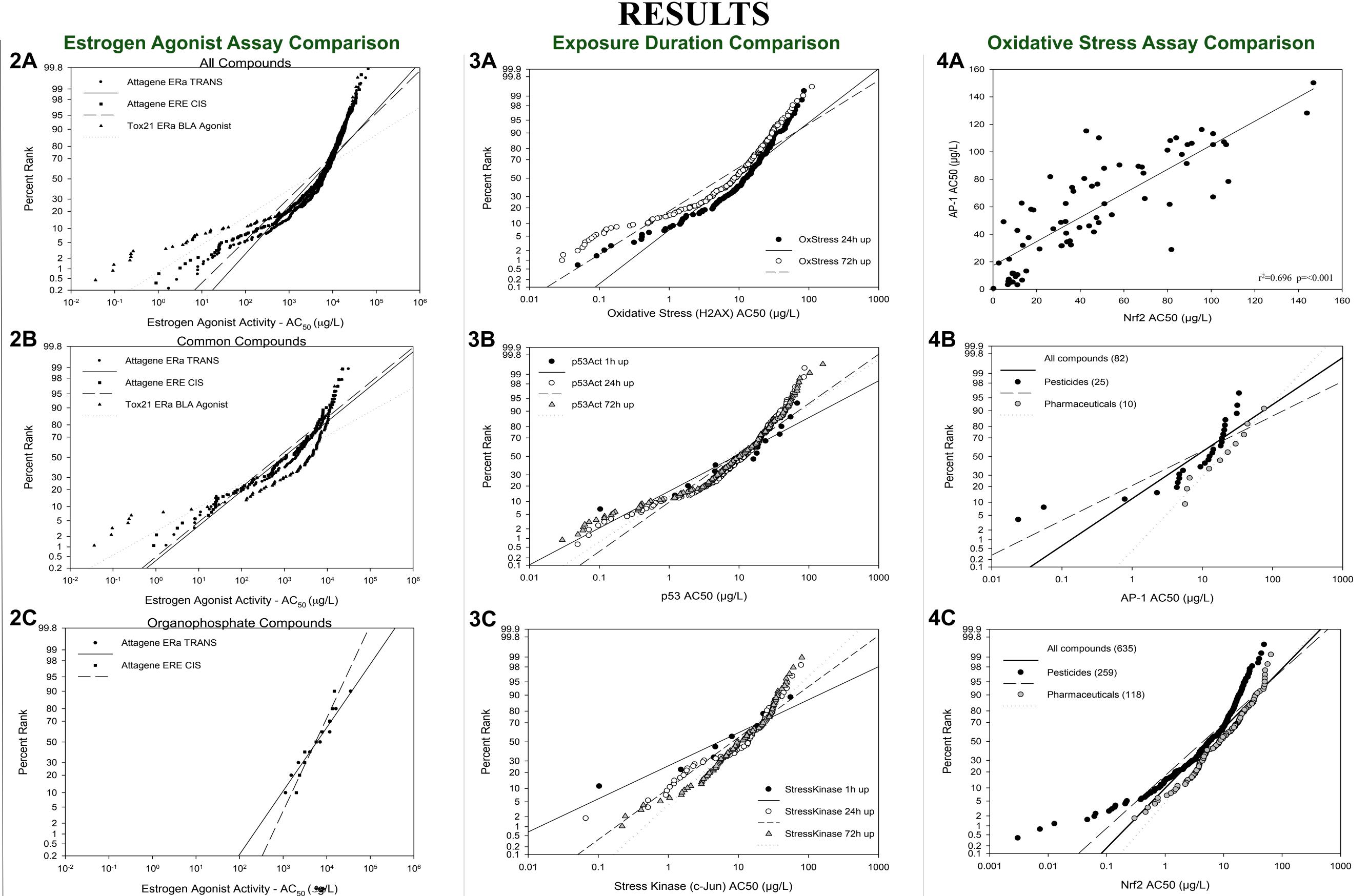


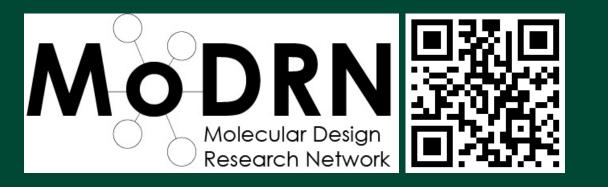
Figure 2: A CTDs for all compounds examined by three in vitro assays of estrogenicity. **B** CTDs for common compounds examined by the three *in vitro* assays. **C** CTDs for common organophosphate compounds examined by two Attagene *in vitro* assays of estrogenicity.

Figure 3: Comparisons between assay duration for A Oxidative Stress (H2AX), **B** p53, and **C** Stress Kinase (c-Jun) assays.

	Equations for regression lines and CTD co Assay	n	$r^2$	a (slope)	b (intercept)	Centile value (µg/L)	
						1%	5%
[2A]	Attagene_Era_TRANS	437	0.859	1.238	-4.417	48.8	173
	Attagene_ERE_CIS	282	0.858	1.097	-3.789	21.5	90.0
	Tox21 ERa BLA Agonist	227	0.724	0.710	-2.400	1.27	11.6
[2B]	Attagene_Era_TRANS	90	0.924	0.903	-2.682	2.48	14.1
	Attagene_ERE_CIS	90	0.939	0.906	-2.587	1.94	11.0
	Tox21 ERa BLA Agonist	90	0.734	0.621	-1.911	0.214	2.69
[2C]	Attagene_Era_TRANS	9	0.975	1.602	-6.045	210	558
	Attagene_ERE_CIS	9	0.941	2.349	-8.797	568	1110
	Assay	n	$r^2$	a	b	Centile value (µg/L)	
				(slope)	(intercept)	1%	5%
[3A]	Oxidative Stress (H2AX) 24hr	147	0.897	1.525	-1.472	0.275	0.77
	Oxidative Stress (H2AX) 72hr	205	0.883	1.234	-0.930	0.074	0.263
[3B]	p53 activity 1hr	14	0.890	1.044	-0.987	0.052	0.234
	p53 activity 24hr	160	0.895	1.398	-1.299	0.184	0.565
	P53 activity 72hr	223	0.901	1.294	-1.129	0.119	0.399
[3C]	Stress Kinase (c-Jun) 1hr	8	0.900	0.907	-0.659	0.014	0.082
	Stress Kinase (c-Jun) 24hr	55	0.916	1.397	-1.282	0.179	0.550
	Stress Kinase (c-Jun) 72hr	96	0.945	1.759	-1.717	0.450	1.099
[4A]	Nrf2 vs AP-1	73	0.696	0.876	17.193		
[4B]	<b>AP-1</b>	82	0.852	1.315	-1.179	0.134	0.442
	Pesticides	25	0.716	0.972	-0.814	0.028	0.140
	Pharmaceuticals	10	0.962	2.064	-2.622	1.391	2.975
[4C]	Nrf2	635	0.899	1.648	-1.287	0.234	0.607
	Pesticides	259	0.887	1.413	-1.145	0.146	0.443
	Pharmaceuticals	118	0.952	2.126	-1.670	0.491	1.028

# Probabilistic Environmental Hazard Assessments of ToxCast Phase I and II In Vitro Datasets

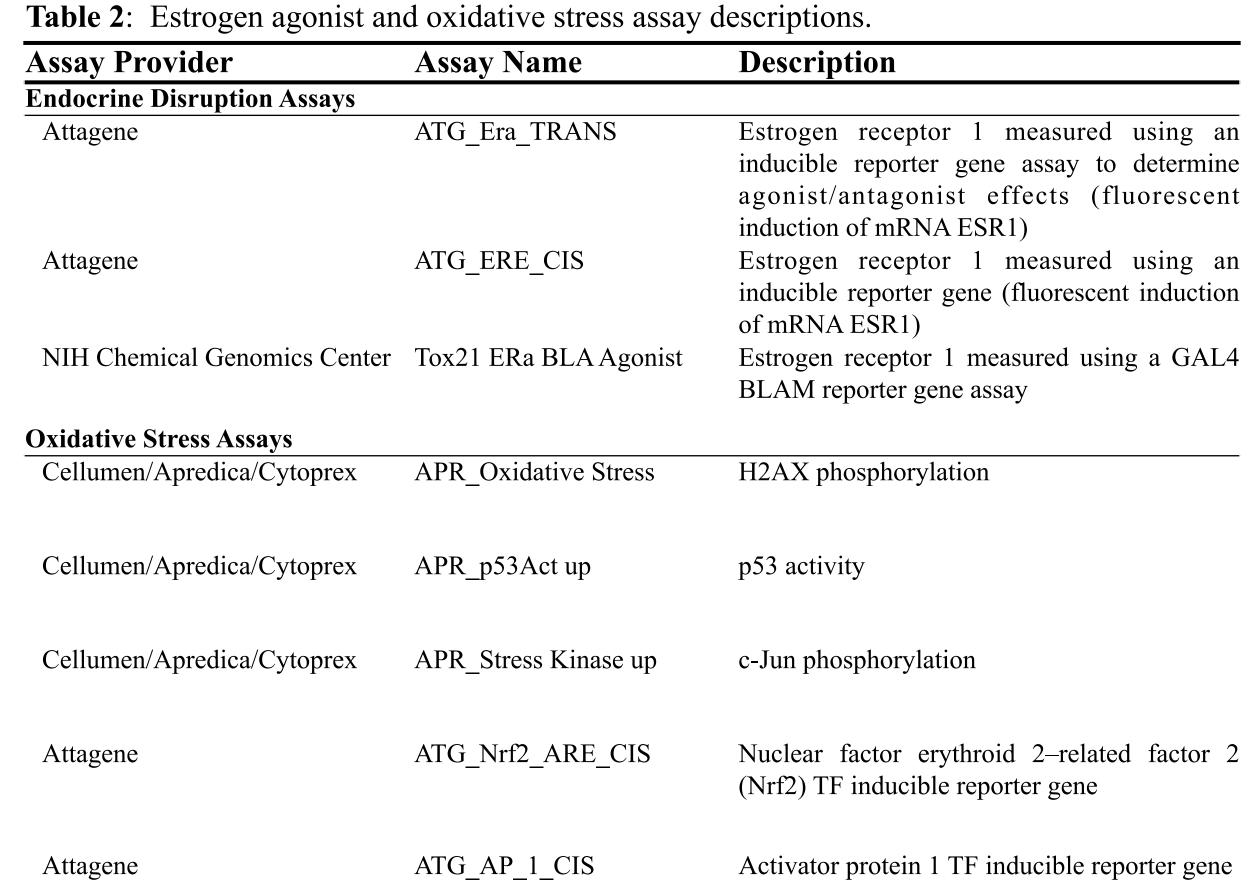
Figure 4: A Regression comparison between common compounds (72) in the AP-1 and Nrf2 assays. **B** CTD for the AP-1 assay with specific chemical compound class regressions. C CTD for the Nrf2 assay with specific chemical compound class regression comparisons.



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# DISCUSSION

- Estrogen agonist in vitro assays from ToxCast vary in their relative sensitivities; the most sensitive assay varied when all data, common compounds or organophosphate compounds were considered. NCGC Agonist assay was most sensitive when all compounds and common chemicals were considered, but the Attagene TRANS assay was most sensitive when organophosphates were compared, with 5<sup>th</sup> centile values of 11.6, 2.69, and 558  $\mu$ g/L, respectively. (Table 1; Figure 2A-C)
- Exposure duration altered assay sensitivity; for the oxidative stress (H2AX) assay sensitivity increased with duration (Figure 3A), while for both the p53 and Stress Kinase (c-Jun) assays sensitivity decreased with duration (Figure 3B-C).
- High correlation was seen between the AP-1 and Nrf2 assays ( $r^2=0.696$ ) (Table 1; Figure 4A). Both assays were more sensitive to pesticides (0.253, 0.443) than pharmaceuticals (2.012, 1.028) as can be seen by their 5<sup>th</sup> centile values, respectively (Table 1; Figure 4B-C).
- Our findings suggest assay selection for estrogenicity and oxidative stress should consider false negatives from ToxCast, chemical class and experimental assumptions.

## **NEXT STEPS**

- Identify physico-chemical properties of compounds with a wide range and degree of affecting oxidative stress-regulated genes
- Develop a computational model that identifies attributes of less hazardous chemicals and elucidates sustainable molecular design guidelines to minimize health effects of oxidative stress
- Conduct *in vitro* and *in vivo* exposures to evaluate activation of oxidative stress-regulated genes and thus strengthen, expand, and confirm computational modeling

## REFERENCES

1. Berninger JP, Brooks BW. Leveraging mammalian pharmaceutical toxicology and pharmacology data to predict chronic fish responses to pharmaceuticals. *Toxicology Letters* 2010;193(1):69-78. 2. Dobbins LL, Brain RA, Brooks BW. Comparison of the sensitivities of common in vitro and in vivo assays of estrogenic activity: Application of chemical toxicity distributions. Environmental Toxicology and Chemistry 2008;27(12) 2608-16.

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