

Southern California Salinity Coalition

TDS Trends Study

Update for the Basin Monitoring Program Task Force

February 14, 2018



Acknowledgements

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- Data provided by:
 - City of Riverside Public Utilities
 - City of San Bernardino
 - Eastern Municipal Water District
 - Inland Empire Utilities Agency
 - Los Angeles Sanitation Districts
 - Orange County Water District / Orange County Sanitation District
 - San Diego County Water Authority
 - Metropolitan Water District of Southern California
- Technical Direction - Risk Sciences



TDS Trends Study - Synopsis

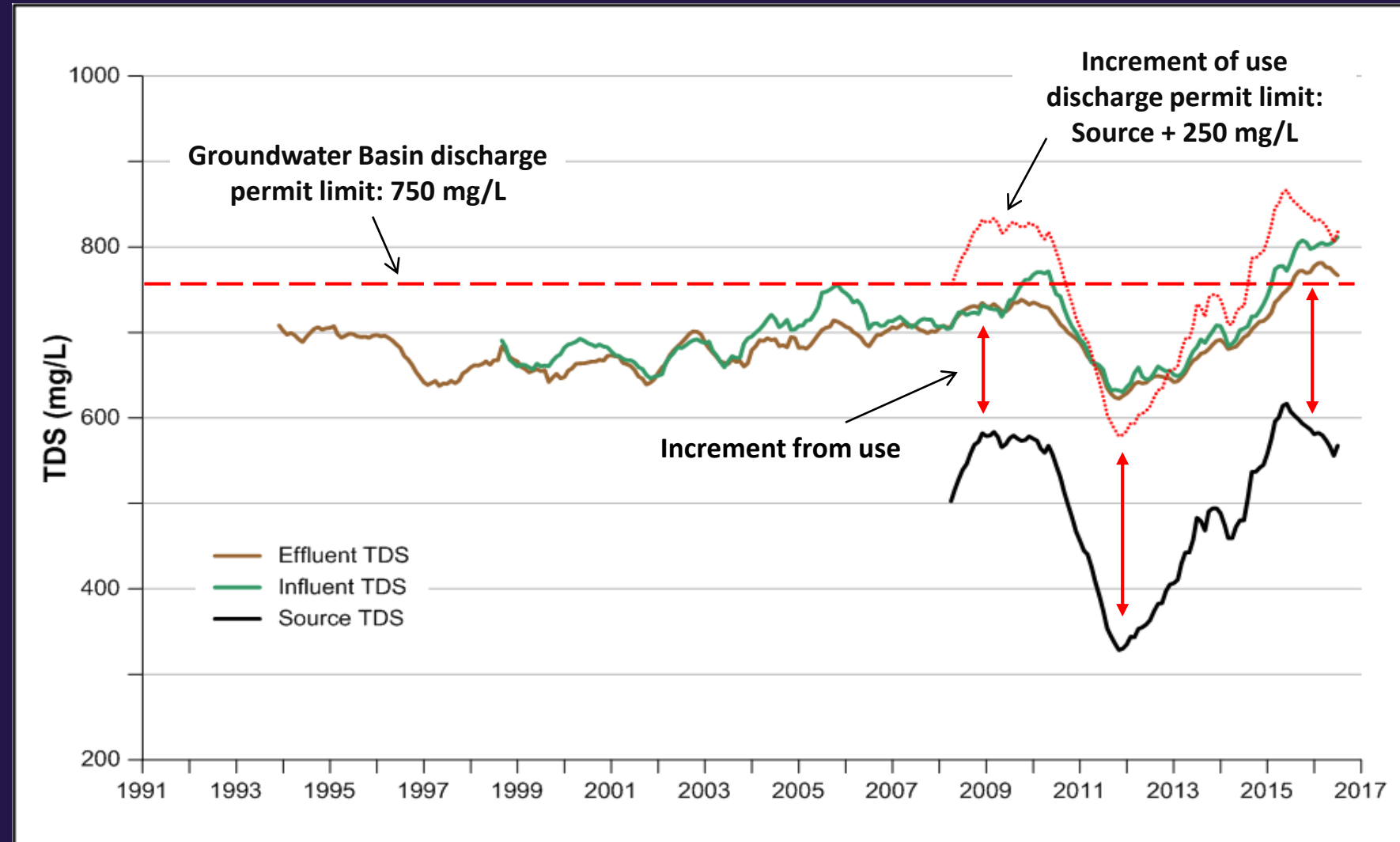
- Identify the effects of drought and water conservation measures on the long-term TDS trends in wastewater and recycled water
- Drought, water conservation measures, and other explanatory variables are intertwined (auto-correlated) to some degree
- Study analyzed both deterministic models and statistical models (multiple linear regression) to predict TDS in wastewater and recycled water
- Provide the science and statistical analysis to provide a framework for policy discussions



TDS Trends Example - Temecula Valley WRF

Considerations:

- 12-mo average period
- Influent ~ Effluent
- Discharge limit based on IFU limit and absolute limits.



Multiple Linear Regression: Influent TDS

$$y_i = b_0 + \sum_{j=1}^n b_j x_{ij} + e_i$$

where

y_i = the predicted value of the response variable y for data point i

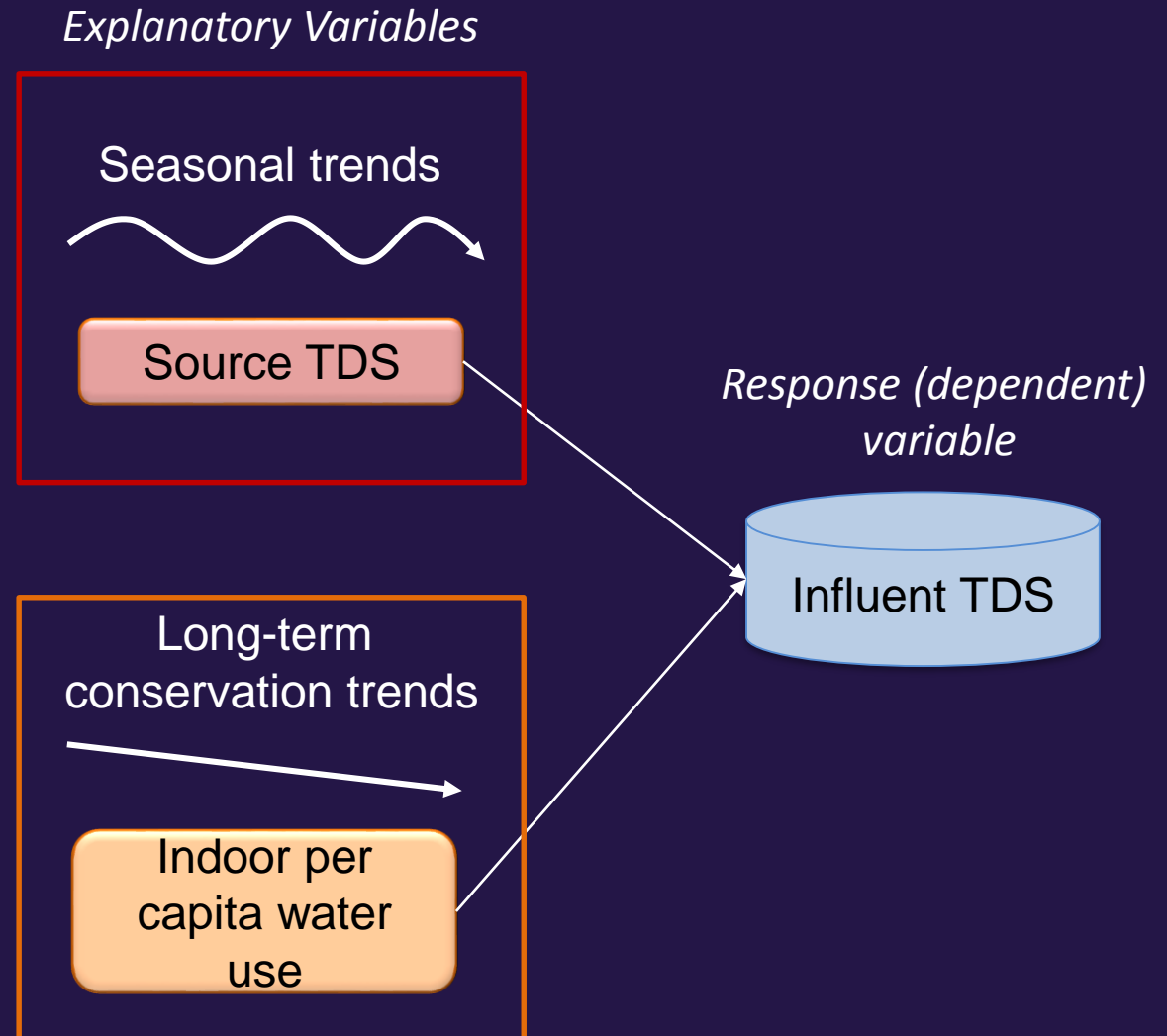
b_0 = the model intercept coefficient

b_j = the model slope coefficient for explanatory variable j

n = the total number of explanatory variables in the model

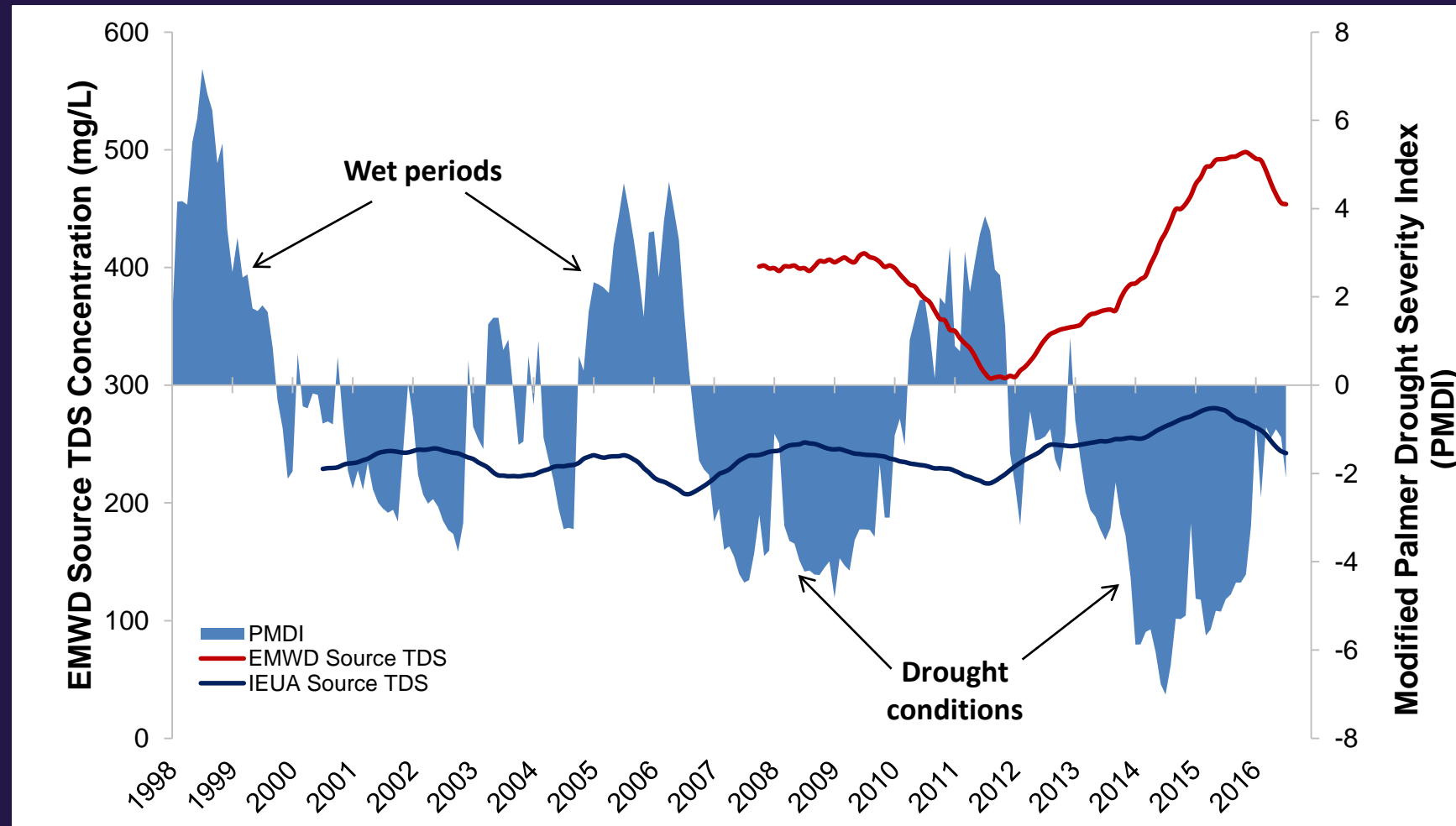
x_{ij} = the known value x of explanatory variable j for data point i

e_i = the residual error of data point i from the fitted model

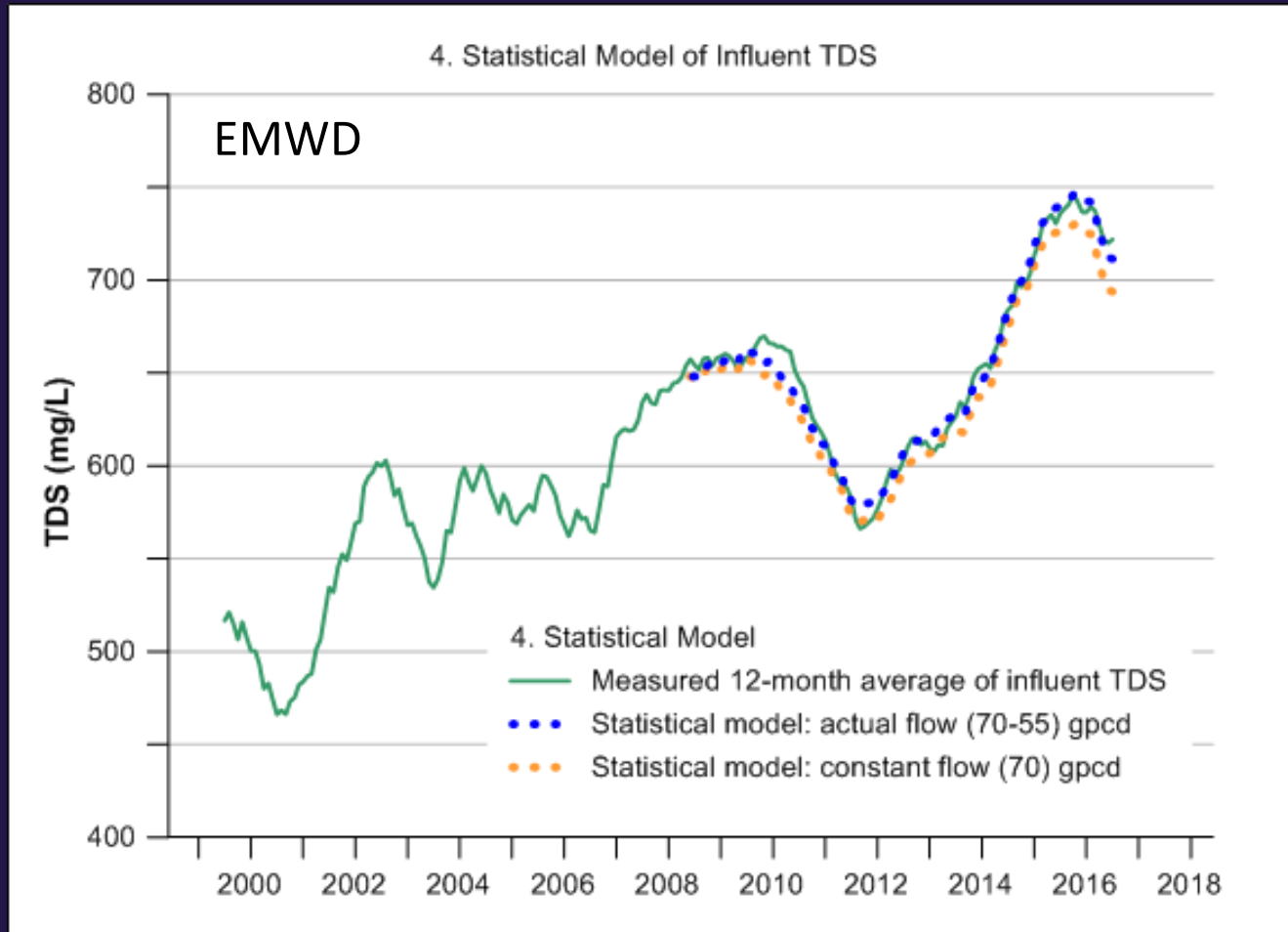


Source Supply TDS Concentrations and Drought

- Higher TDS concentration with drought periods
- EMWD greater reliance on imported water
- IEUA greater reliance on groundwater and local water supply

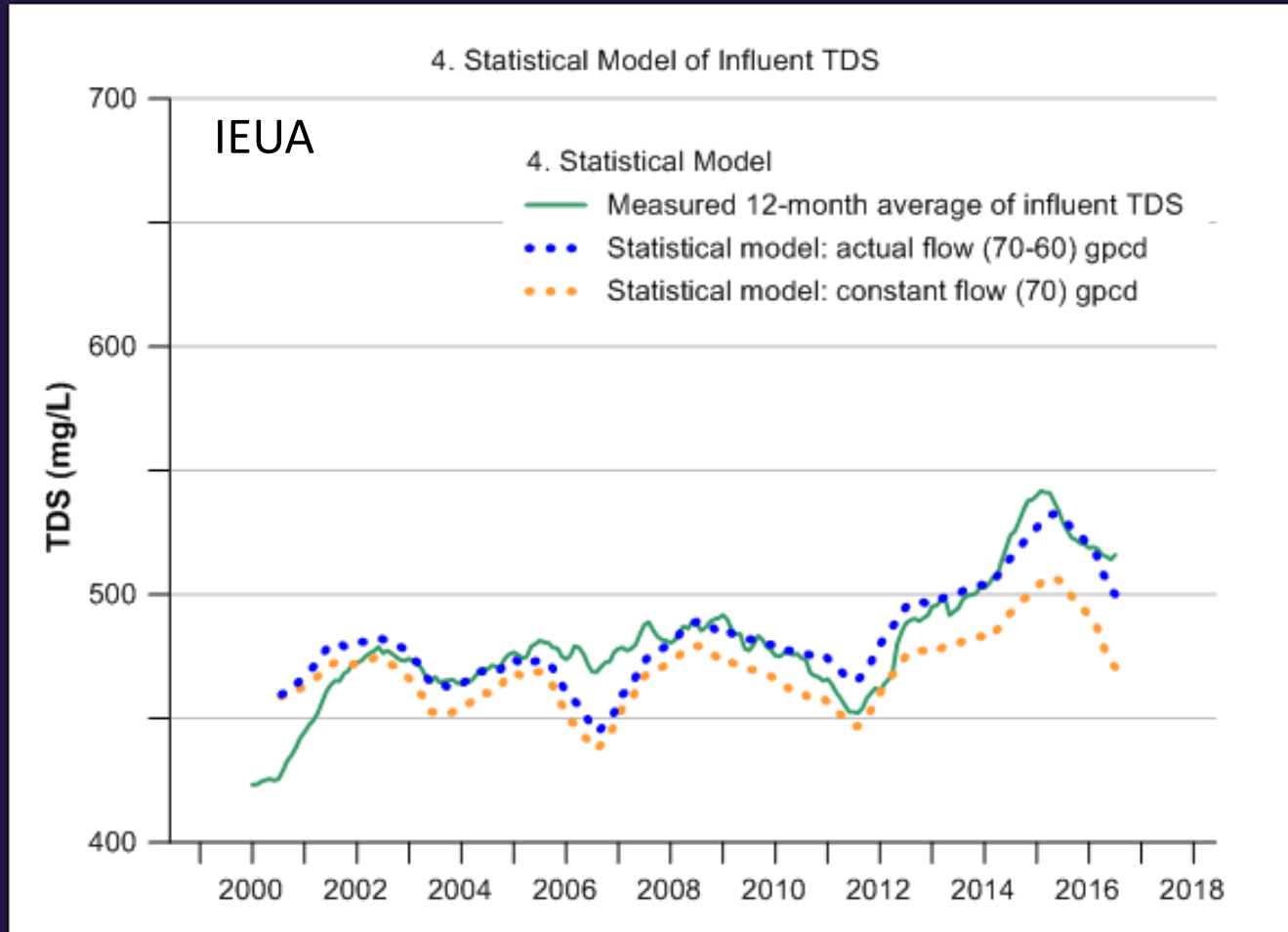


Multiple Linear Regression: Influent TDS



- Variables:
 - STDS: Source TDS
 - IGPCD: Influent per capita water use
- R-squared = 0.98
- Relative Importance (%)
 - STDS: 88.2
 - IGPCD: 11.8

Multiple Linear Regression: Influent TDS



- Variables:
 - STDS: Source TDS
 - IGPCD: Influent per capita water use
- R-squared = 0.75
- Relative Importance (%)
 - STDS: 67.2
 - IGPCD: 32.8

TDS Statistical Model Matrix

- Using the statistical models, matrices were developed to predict the effects of conservation and changes in source water TDS. Much of this variation was due to climatic factors such as drought.
- EMWD Example: During the peak of the drought, source water quality was approximately 500 mg/L and indoor per capita water use was 55 gpcd. The estimated water quality entering a WWTP would be approximately 750 mg/L.



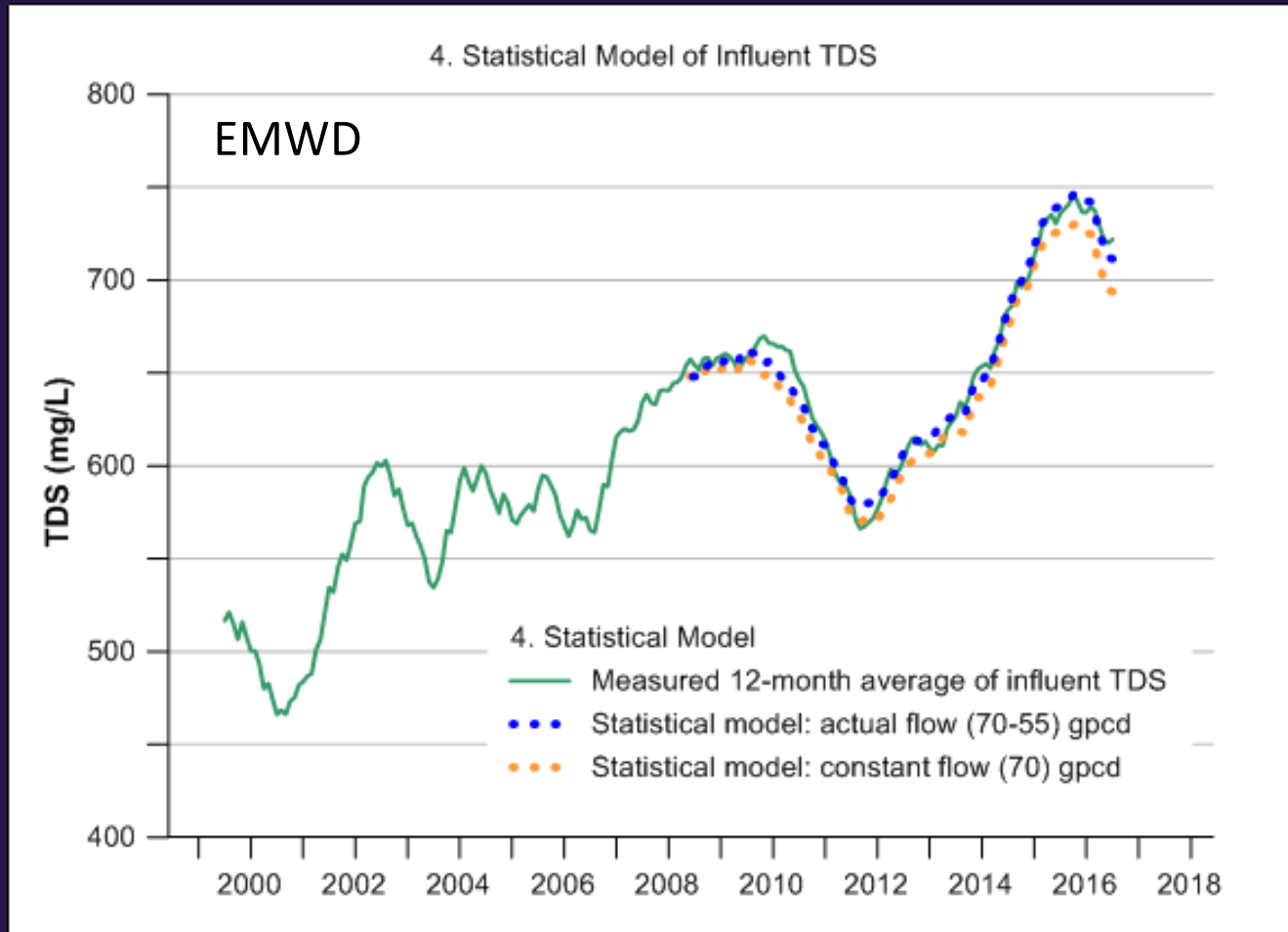
EMWD Statistical Model Matrix for Influent TDS

		Source TDS (mg/L)												
		300	325	350	375	400	425	450	475	500	525	550	575	600
Indoor Water Use (gpcd)	40	608	629	650	671	692	713	733	754	775	796	817	838	859
	42	605	626	646	667	688	709	730	751	772	793	814	835	856
	44	601	622	643	664	685	706	727	748	769	790	810	831	852
	46	598	619	640	661	682	703	724	744	765	786	807	828	849
	48	595	616	637	657	678	699	720	741	762	783	804	825	846
	50	591	612	633	654	675	696	717	738	759	780	801	821	842
	52	588	609	630	651	672	693	714	735	755	776	797	818	839
	54	585	606	627	648	668	689	710	731	752	773	794	815	836
	56	581	602	623	644	665	686	707	728	749	770	791	812	832
	58	578	599	620	641	662	683	704	725	746	766	787	808	829
	60	575	596	617	638	659	679	700	721	742	763	784	805	826
	62	572	592	613	634	655	676	697	718	739	760	781	802	823
	64	568	589	610	631	652	673	694	715	736	756	777	798	819
	66	565	586	607	628	649	670	690	711	732	753	774	795	816
68	562	583	603	624	645	666	687	708	729	750	771	792	813	
70	558	579	600	621	642	663	684	705	726	747	767	788	809	

Every 1 gpcd decrease amounts to 1.7 mg/L increase in TDS



Multiple Linear Regression: Influent TDS



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Long-term rolling averages

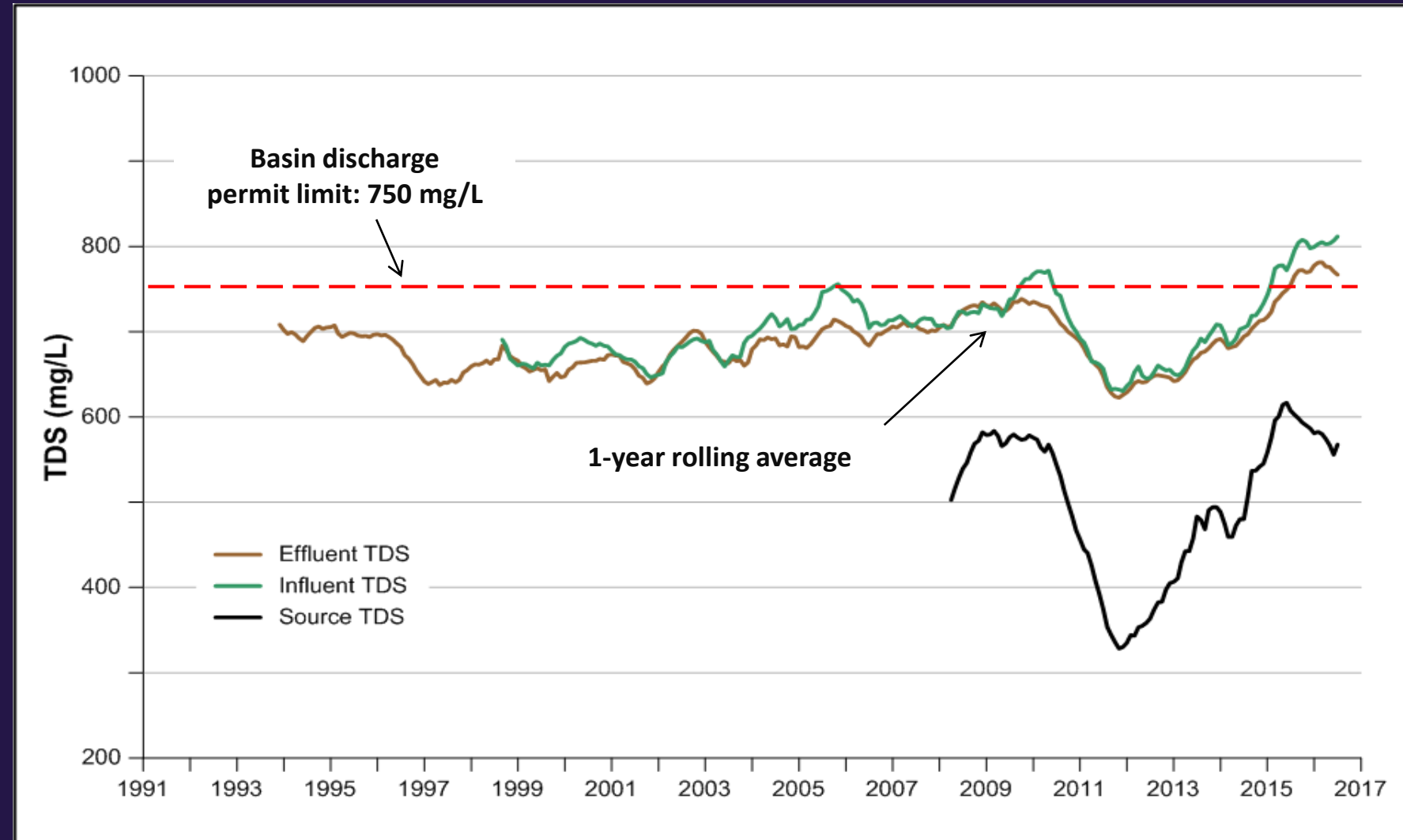
- How does the volume-weighted average TDS concentration in recycled water, and the related increment of use, vary using a range of rolling averaging periods (e.g., 1, 5, 10, and 15 years)?
- Longer-term rolling average periods smooth out annual variations of effluent trends. 10 year averages account for seasonal cyclicity.



TDS Trends Example - Temecula Valley

Considerations:

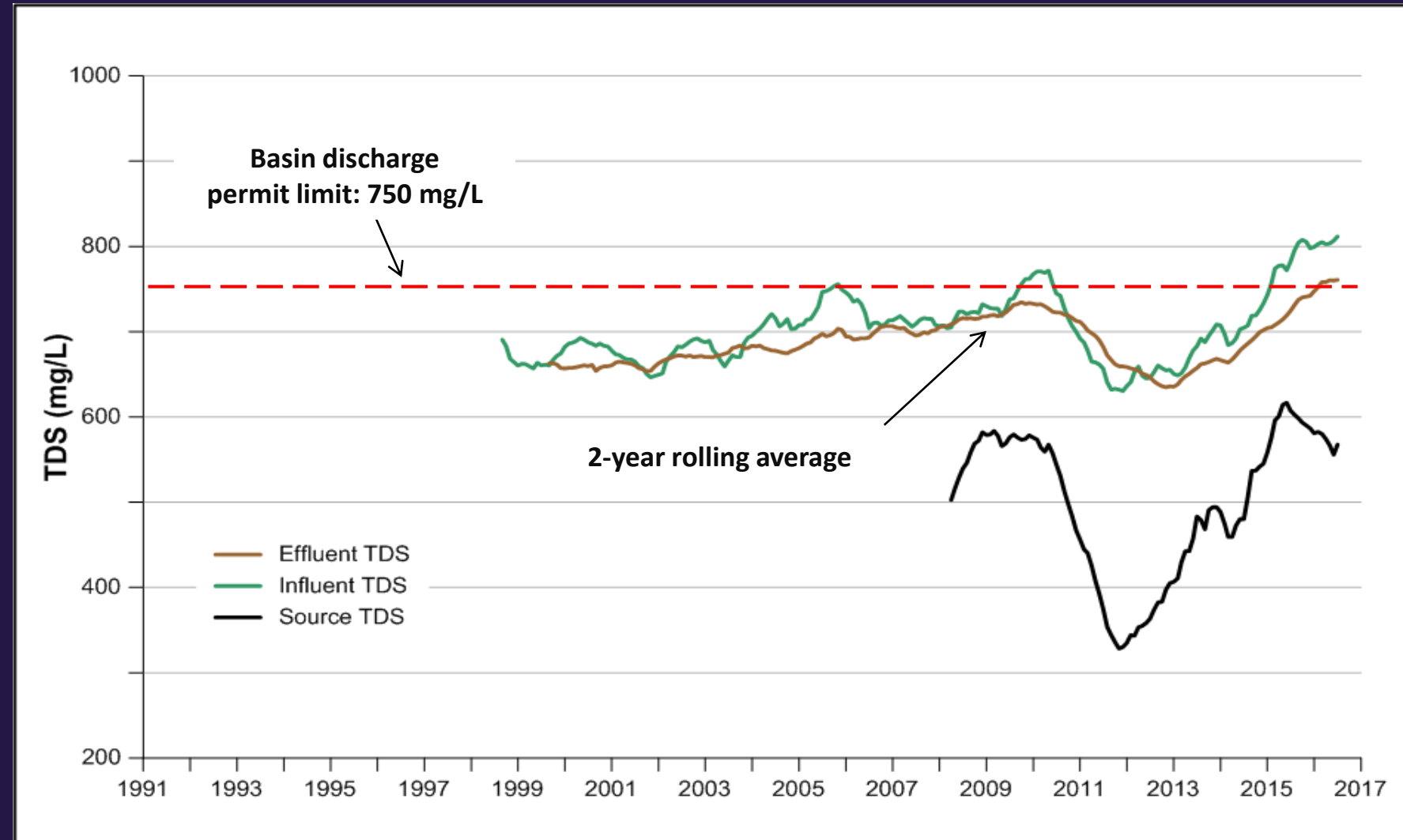
- Rolling average period
- Discharge limits based on Management Zone Water Quality Objectives
- Long term trends
- Seasonal cyclicity (drought vs wet years)



TDS Trends Example - Temecula Valley

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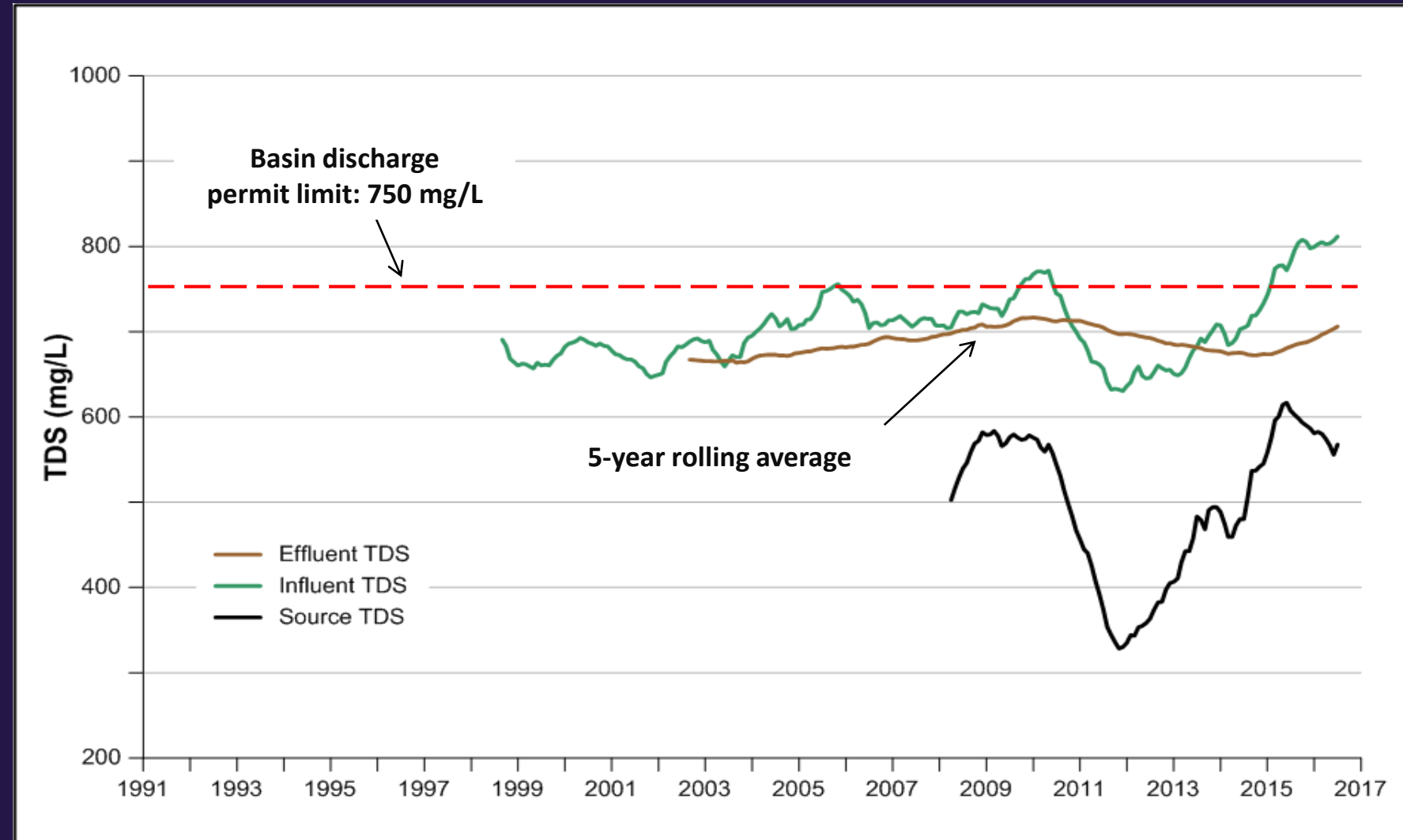
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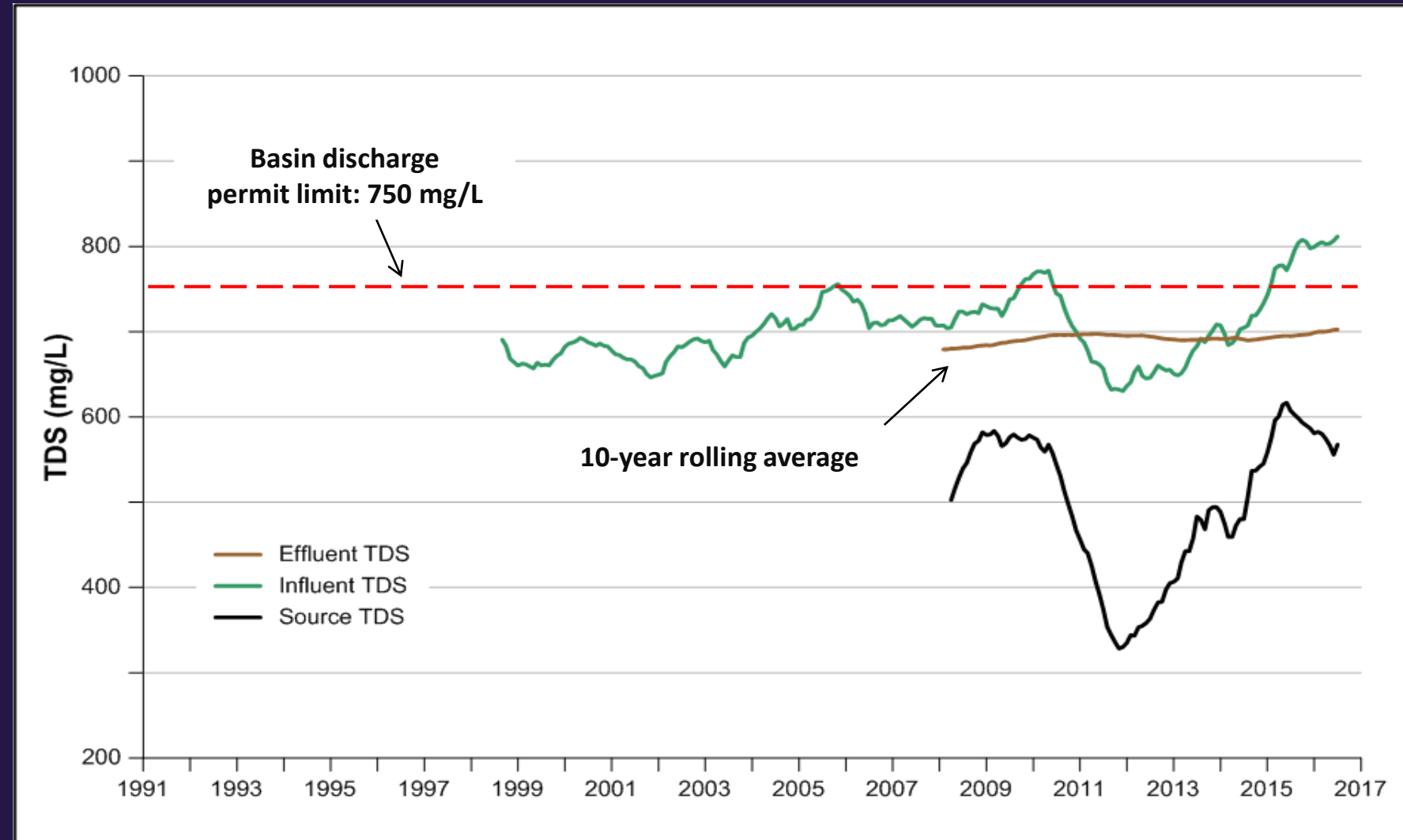
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Summary

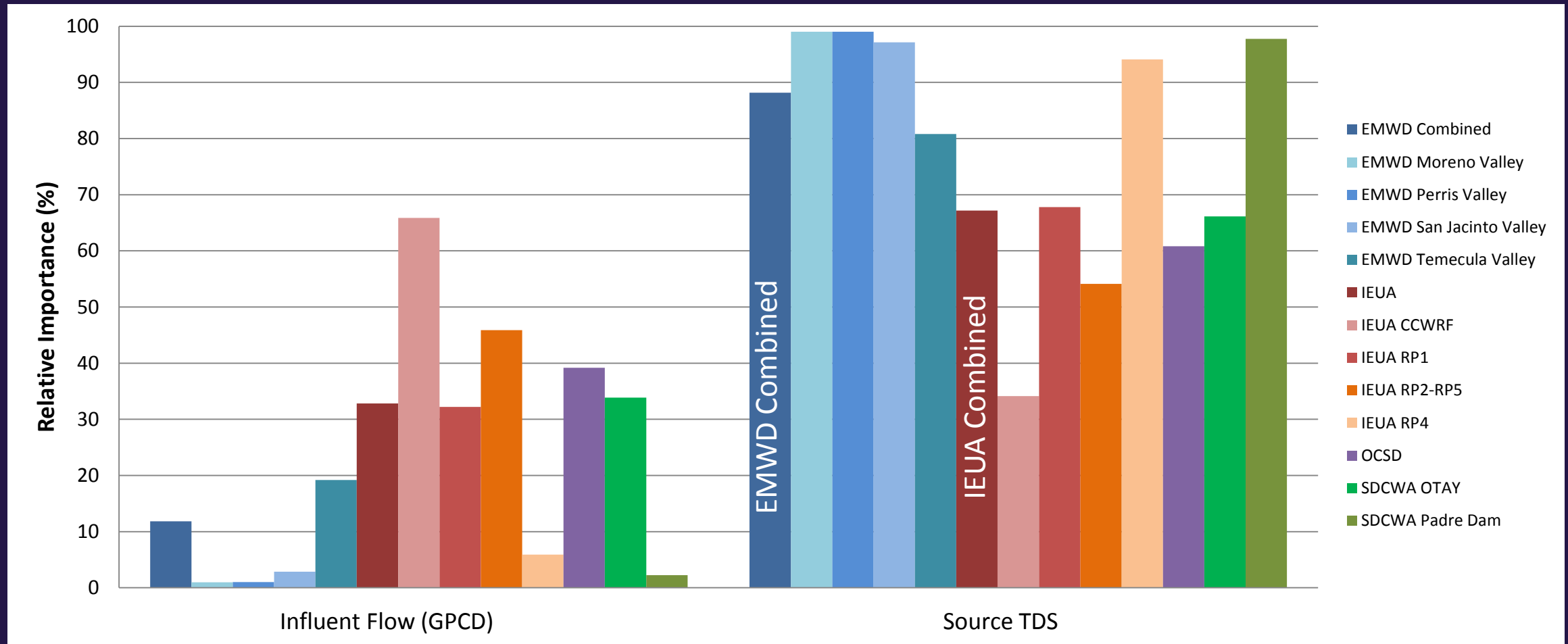
- Longer rolling averages (>5-years) minimize the influence of drought cycles. Long-term upward trends in TDS will still be present.
- Statistical modeling suggests that for every 1.0 gallon per capita per day that is conserved there will be an increase in TDS concentrations to the WWTPs of 1.2 mg/L to 1.7 mg/L
- Unintended consequences from water conservation measures
 - lower water quality (higher TDS)
 - less quantity of recycled water
 - less revenue
 - infrastructure O&M
 - Less energy uses
 - Less GHG emissions



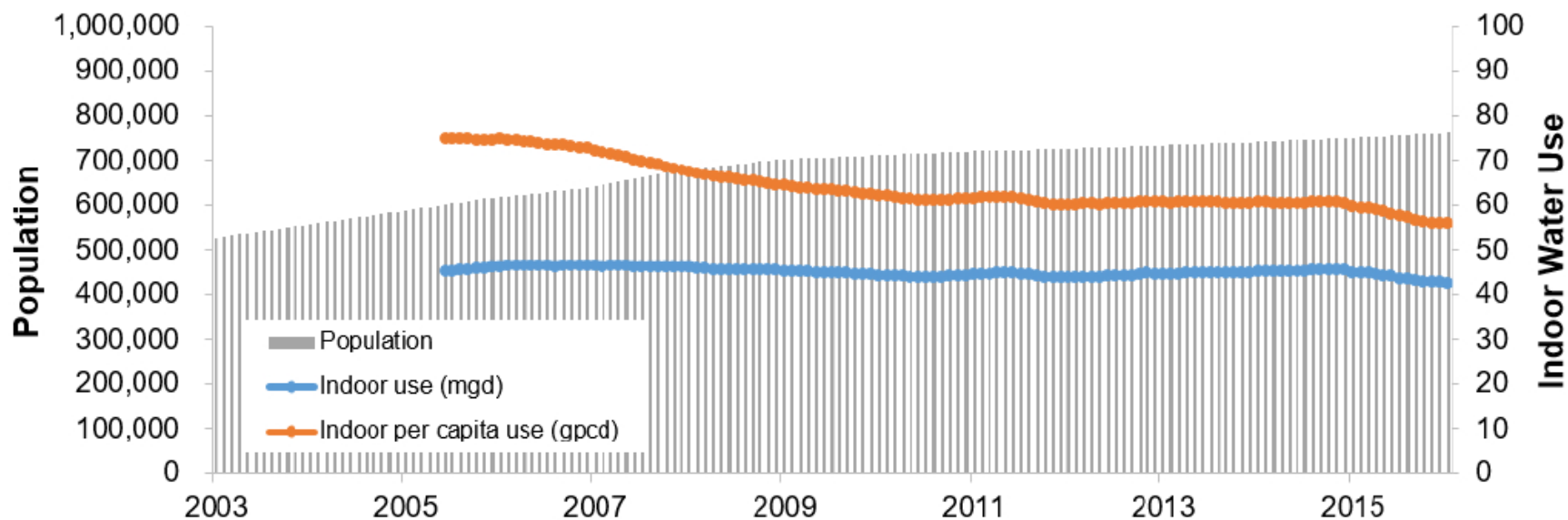
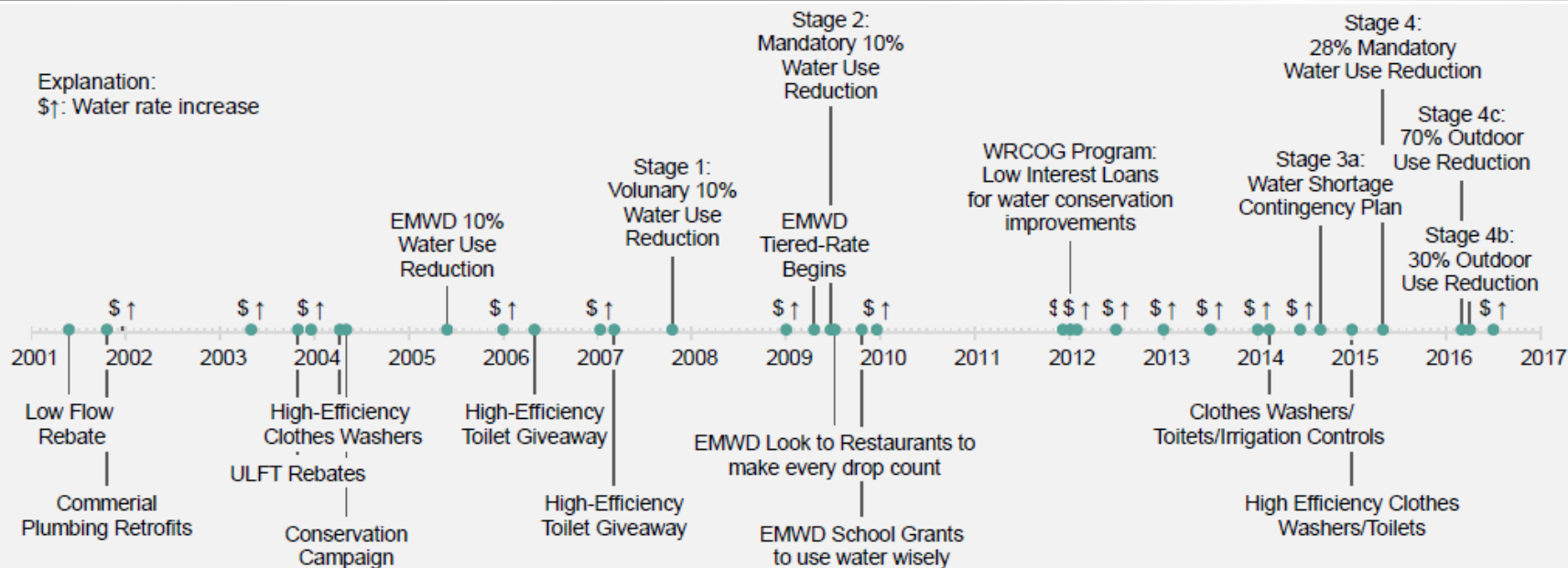
Questions?



MLA Relative Importance



Explanation:
\$↑: Water rate increase



Self-Regenerating Water Softeners

Santa Clarita Valley Sanitation District (SCVSD)

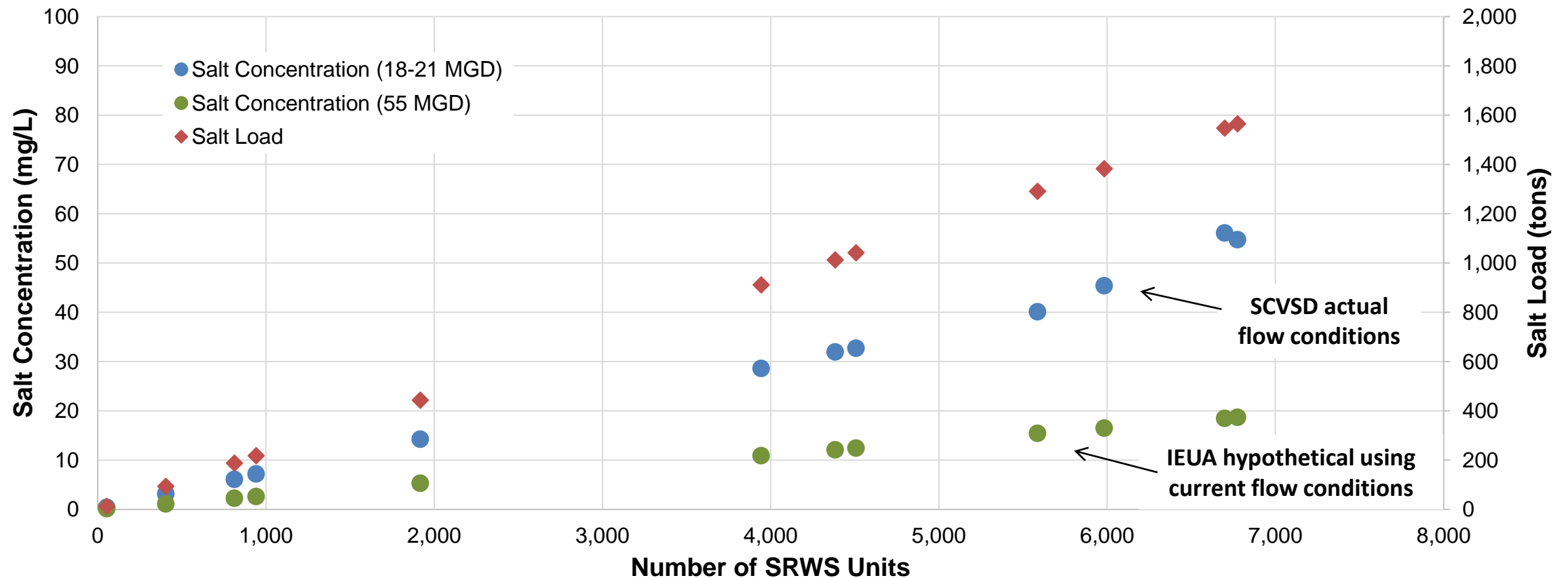
- March 2003 SRWS installation ban ordinance takes effect
- November 2005 Voluntary Phase I Rebate Program
- May 2007 Voluntary Phase II Rebate Program
- January 2009 mandatory ordinance banning SRWS
- August 2011 Ordinance Enforcement Program

Year	Estimate number SRWS
2002	5,983
2003	6,699
2004	6,775
2005	5,587
2006	4,384
2007	4,507
2008	3,943
2009	1,917
2010	812
2011	942
2012	54

Los Angeles County Sanitation District Chloride Study (2014)



Self-Regenerating Water Softeners



Summary

- Reduction of SRWS can significantly reduce the concentration of TDS to WWTPs if there are enough removed from the system. By removing 6,000 SRWS units, it is estimated that SCVSD reduced the TDS brine contribution to the wastewater influent flow for that agency by 50 mg/L. Removing the same number of SRWS from IEUA could reduce the concentration by 17 mg/L.



Summary

- Observation data from groups of sewerage agencies rather than individual WWTP is more reliable due to the following factors
 - Population (city boundaries, sewerage boundaries)
 - Operations can divert flows from plant to plant
- Source TDS in combination with indoor per capita water use can predict the influent TDS to WWTPs with high levels of certainty
- Drought conditions negatively impact surface water quality and therefore source water quality and will become increasingly important if drought cycle patterns intensify due to climate change

