



# Drinking water treatment plant (DWTP) costs and source water quality: An updated case study

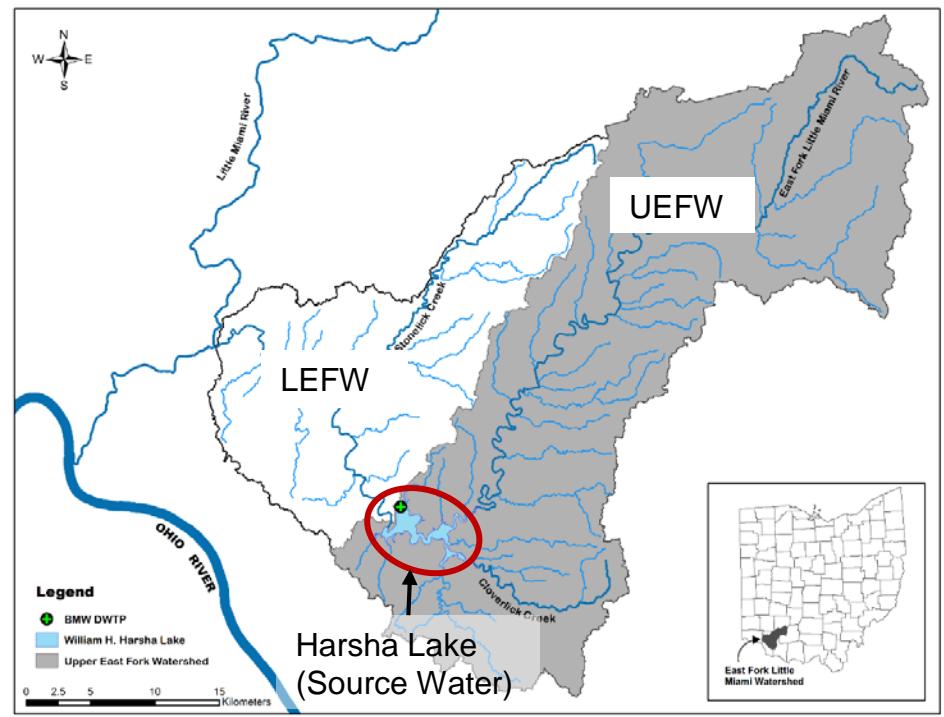
Matthew T. Heberling, Christopher T. Nietch, James I. Price,  
Hale W. Thurston, Michael Elovitz

US Environmental Protection Agency, Office of Research and Development, Cincinnati, OH

Disclaimer: The views expressed in this presentation are those of the authors and do not necessarily represent the views or policies of the US Environmental Protection Agency (EPA). Mention of trade names or commercial products does not constitute EPA endorsement or recommendations for use.

# Background

- **Heberling et al. (2015)**
  - Developed a framework to compare costs:
    - **Drinking water treatment to source water protection**
  - Examine treatment costs with changes in source water quality, production, and seasonal variables
    - Five years daily data for one plant in southwest OH
  - Presents steps to estimate the incentives for source water protection vs. treatment on-site



# General Steps for DWTPs

Step 1: Link changes in source water quality to changes in treatment costs

- Develop a cost function using data on raw water quality variables and treatment operations

Step 2: Link source water quality to watershed load reductions (through land use management)

- Connect watershed variables (e.g., phosphorus load) to source water quality variables governing treatment cost function (e.g., turbidity or total organic carbon [TOC])
- When treatment variables differ from watershed variables a “translation” is needed

Step 3: Estimate costs of the land use management (e.g., cover crops) that leads to watershed load reductions



*Drinking Water Plant Intake Structure on Lake Harsha*

# Case Study: Bob McEwen Water Treatment Plant (BMWTP)

- **Period 1 (Heberling et al. 2015)**

- Before the period of significant source water quality degradation from Harmful Algal Blooms (HABs)
- Clermont County Water Resources Dept. (operator logs, paper records, invoices)
- US Army Corps. (reservoir characteristics)
- Time series: 1826 daily observations, 2007-2011

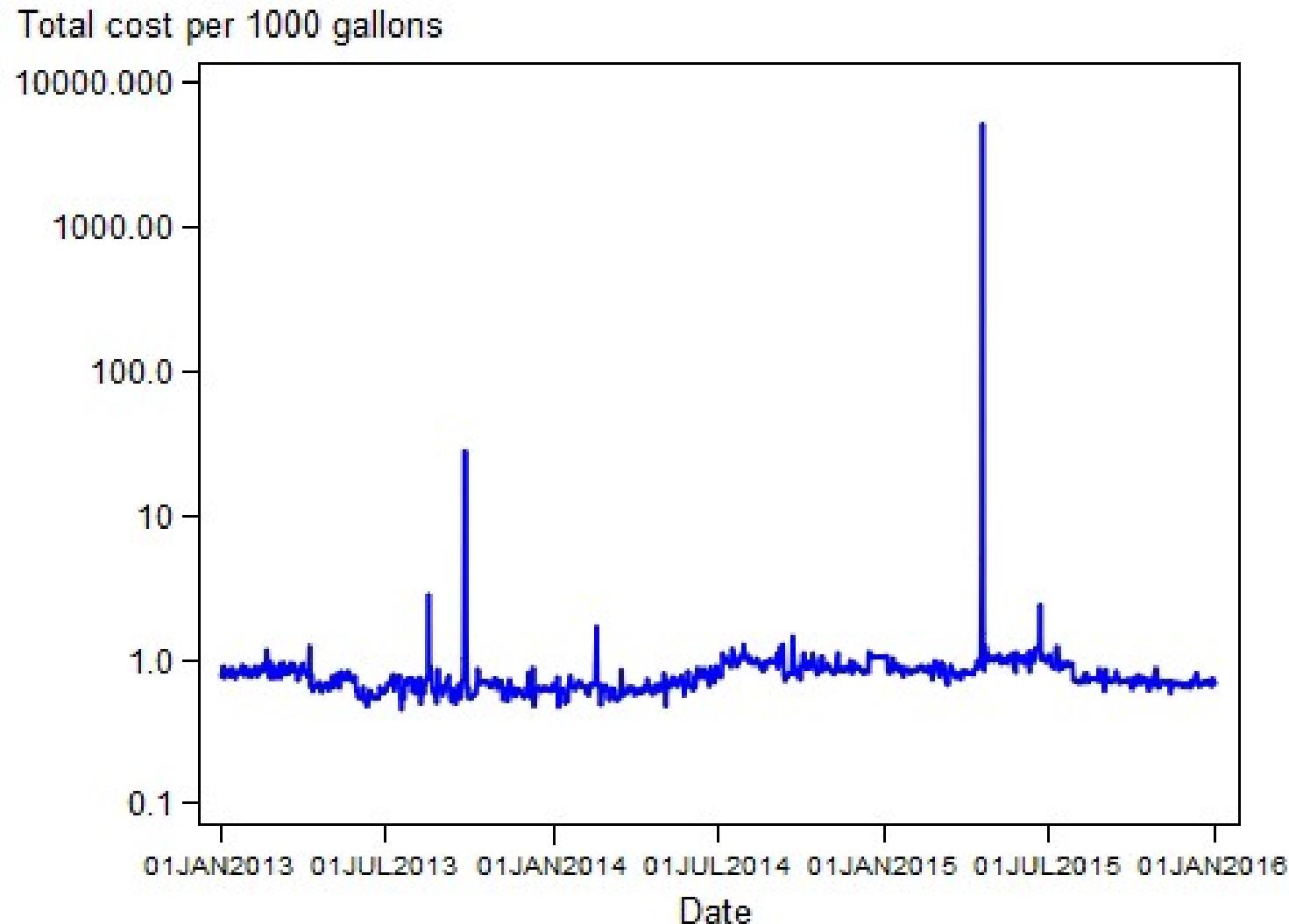
- **Period 2 (Preliminary analysis)**

- In 2012, BMWTP added a granular activated carbon (GAC) building, a significant investment
  - GAC used for addressing disinfection-by-products and algal toxins
  - Greater reliance on GAC filters might translate to increased costs creating need for source protection?
- Same data sources as Period 1
- Time series: 1095 daily observations, 2013-2015

# Summary Statistics

<b>Variable</b>		<b>Period 1</b>	<b>Period 2</b>	<b>Min</b>	<b>Max</b>
		<b>2007-2011</b>	<b>2013-2015</b>		
<b>FINAL</b>	DW produced (MGD)	4.95	3.39	0.001	6.58
<b>POOL</b>	Pool elevation (ft)	731.73	731.90	728.90	752.67
<b>GCPOOL</b>	Pool minus guide curve	0.72	0.89	-1.85	23.67
<b>pH</b>		7.43	7.53	6.52	9.80
<b>TURB</b>	Turbidity (NTU)	11.42	6.72	0.07	73.00
<b>RAWTOC</b>	Raw water TOC (mg/L)	6.10	6.25	4.50	8.30
<b>TOTALCOST/1000</b>	Total costs per 1000 gal	0.55	5.51	0.45	5148.82
<b>ChemCost/1000</b>	Chem cost per 1000 gal	0.23	2.27	0.08	2272.19
<b>PumpCost/1000</b>	Pump cost per 1000 gal	0.23	0.79	0.07	636.65
<b>GACCost/1000</b>	GAC cost per 1000 gal	0.10	2.45	0.16	2239.97

# Total Cost per 1000 Gallons of Final Flow (Dec 2016\$)



# Time Series Model

- Analyze daily costs using Error Correction Model (ECM)
  - Several commonly used time series models ignore the long-run equilibrium relationships predicted by economic theory
  - Tests for long-run and short-run effects
- Changes in one independent variable can have immediate effect on treatment costs, a long-run effect, or both
- Period 2 Model (2013-2015)

$$\frac{Cost}{1000 \text{ gallons}} = f \left( \begin{array}{l} FINAL, RAWTOC, TURB, pH, \\ GCPOOL, \\ ActualTOC, \\ SPR, CY13, CY14, two \text{ } PROCESS \end{array} \right)$$

# ECM Results, Period 2

- Preliminary results
  - RAWTOC
    - 1% **decrease** in RAWTOC ( $\approx 0.06 \text{ mg/L}$ ) leads to an immediate 0.13% decrease in treatment cost
    - No change over future time
  - pH
    - 1% **decrease** in pH ( $\approx 0.08$ ) leads to no immediate effect, but decreases treatment costs by 2.92% over future time
    - Spring months (April, May, June) associated with higher costs



# Link Source Water to Watershed

- How does total phosphorus (TP) load effect source (raw) water?
  - Chose TP load because its linked to:
    1. Hillslope erosion (affinity for clay particles-> sediment load->turbidity)
    2. Algal blooms (increase TOC and pH)
- Time series analysis (polynomial distributed lag [PDL] model)
  - Period 2 Model (2009-2013)
$$RAWTOC = f \left( \begin{matrix} T P L O A D, G C P O O L, S P R, S U M, \\ C Y 0 9, C Y 1 0, C Y 1 1, C Y 1 2 \end{matrix} \right)$$
  - Need better translation among TP load, Algae, and Treatment Process

- Preliminary results (2009-2013)

- 1% reduction in TPLOAD ( $\approx 10.79$  lbs/day) leads to 0.04% decrease in RAWTOC over the long-run (from weeks 2-9)
- \$0.51 treatment cost savings from 1% annual TP load reduction
- Cost to control 10.79 lbs/day for a year using cover crops = \$11815--\$106335 -> Price/lb = \$3—\$27 (i.e.,  $10.79 \times 365 \times \text{price}$ )



# Conclusions

- Framework provides decision support for understanding cost drivers
  - **Period 1:** DWTP could reduce treatment costs in the short and long-run by reduction in source water turbidity from 2007-2011
  - **Period 2:** DWTP could reduce treatment costs by lowering TOC and pH in source water from 2013-2015
    - Cost function better linked to algae problem



## Conclusions (Cont.)

- Periods 1 and 2 revealed no cost incentive to purchase abatement
  - Other DWTPs may have incentive dependent on physiographic setting
  - Statistical model linking watershed to source WQ does not best reflect ecological knowledge
- Limitations:
  - DWTP over-treats due to uncertainties in lake ecology/water quality; hides the impact
  - Not having **appropriate** data for modeling obscures understanding of incentives
  - Linkage between watershed loads and drinking water treatability requires more consideration
  - Waiting for 2017 data to increase time series



# Supplementary Material



## Acknowledgments

- APTIM:
  - Don Schupp
  - Amr Safwat
- Clermont County Water Resources Department:
  - Tim Neyer
  - Mark Day

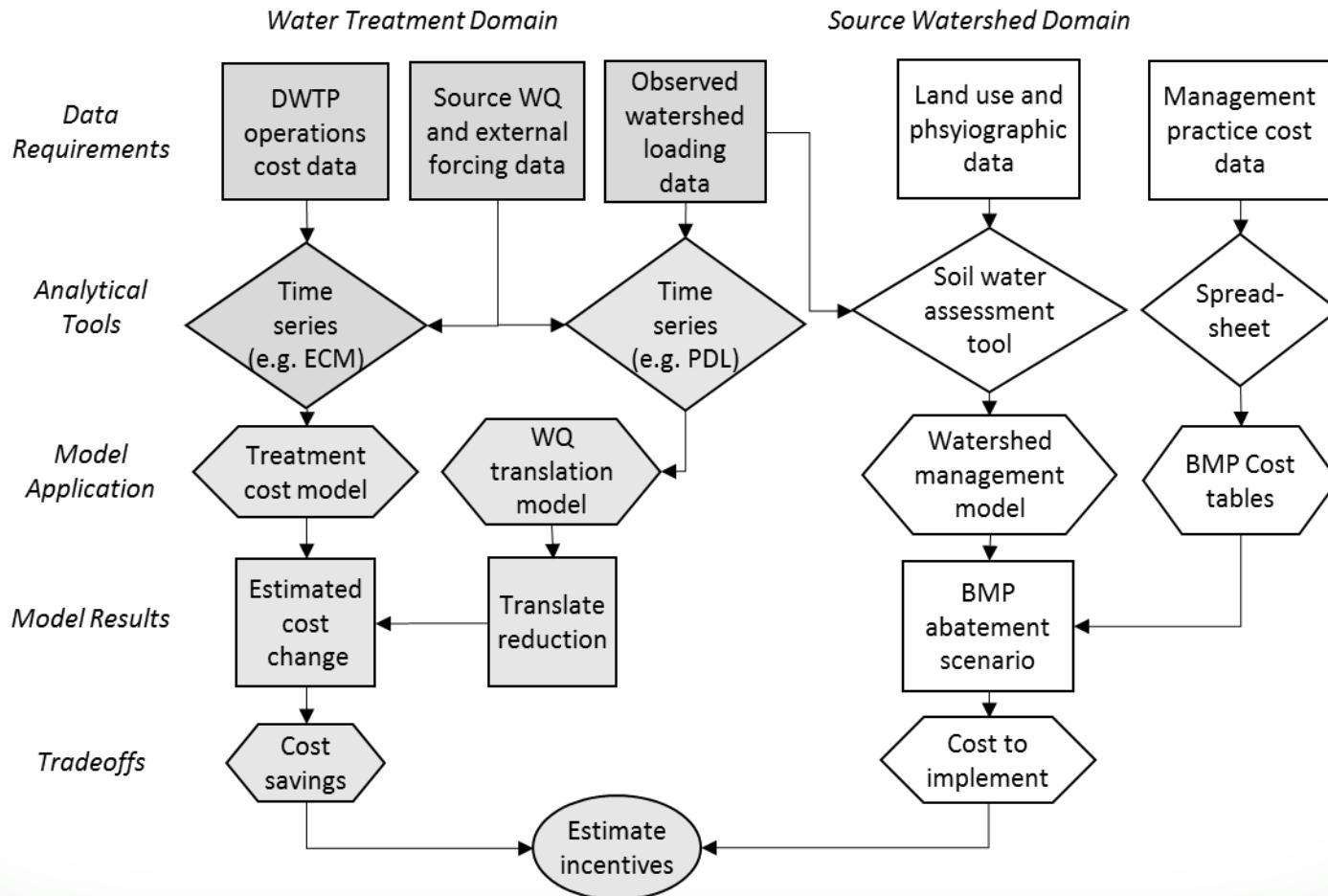
## References

- Gartner, T., J. Mulligan, R. Schmidt, J. Gunn (2013), Natural Infrastructure: Investing in Forested Landscapes for Source Water Protection in the United States, World Resources Institute, Washington, DC.
- Heberling, M., C. Netch, H. Thurston, M. Elovitz, K. Birkenhauer, S. Panguluri, B. Ramakrishnan, E. Heiser, T. Neyer (2015), Comparing Drinking Water Treatment Costs to Source Water Protection Costs Using Time Series Analysis. *Water Resources Research* 51:8741–8756, doi:10.1002/2014WR016422.

# Background

- Concerns about water quality and drinking water safety, especially watershed nutrient loads
- Understanding how treatment costs are affected by changes in source water quality is essential to understanding tradeoffs between source water protection and treatment on site
  - Can provide evidence whether it is less expensive to invest in natural infrastructure than to pay for treatment on site
  - **Important knowledge gap for municipalities and DWTPs** (Gartner et al. 2013)

# Framework



# ECM equation and PDL equation

$$\Delta COST_t = \alpha_0 + \sum_{i=1}^{p-1} \alpha_i \Delta COST_{t-i} + \sum_{j=1}^n \sum_{i=0}^{q-1} \beta_{ji} \Delta X_{jt-i} + \alpha_p \left[ COST_{t-p} - \sum_{j=1}^n \omega_j X_{jt-q} \right] + \varepsilon_t \quad (1)$$

$$\begin{aligned} \ln TOC_t = & \alpha_0 + \beta_0 * GCPool_t + \beta_1 * SPR + \beta_2 * SUM + \beta_3 \\ & * CY09 + \beta_4 * CY10 + \beta_5 * CY11 + \beta_6 * CY12 \\ & + \sum_{i=0}^v \theta_i \ln TPLoad_{t-i} + u_t \end{aligned} \quad (2)$$

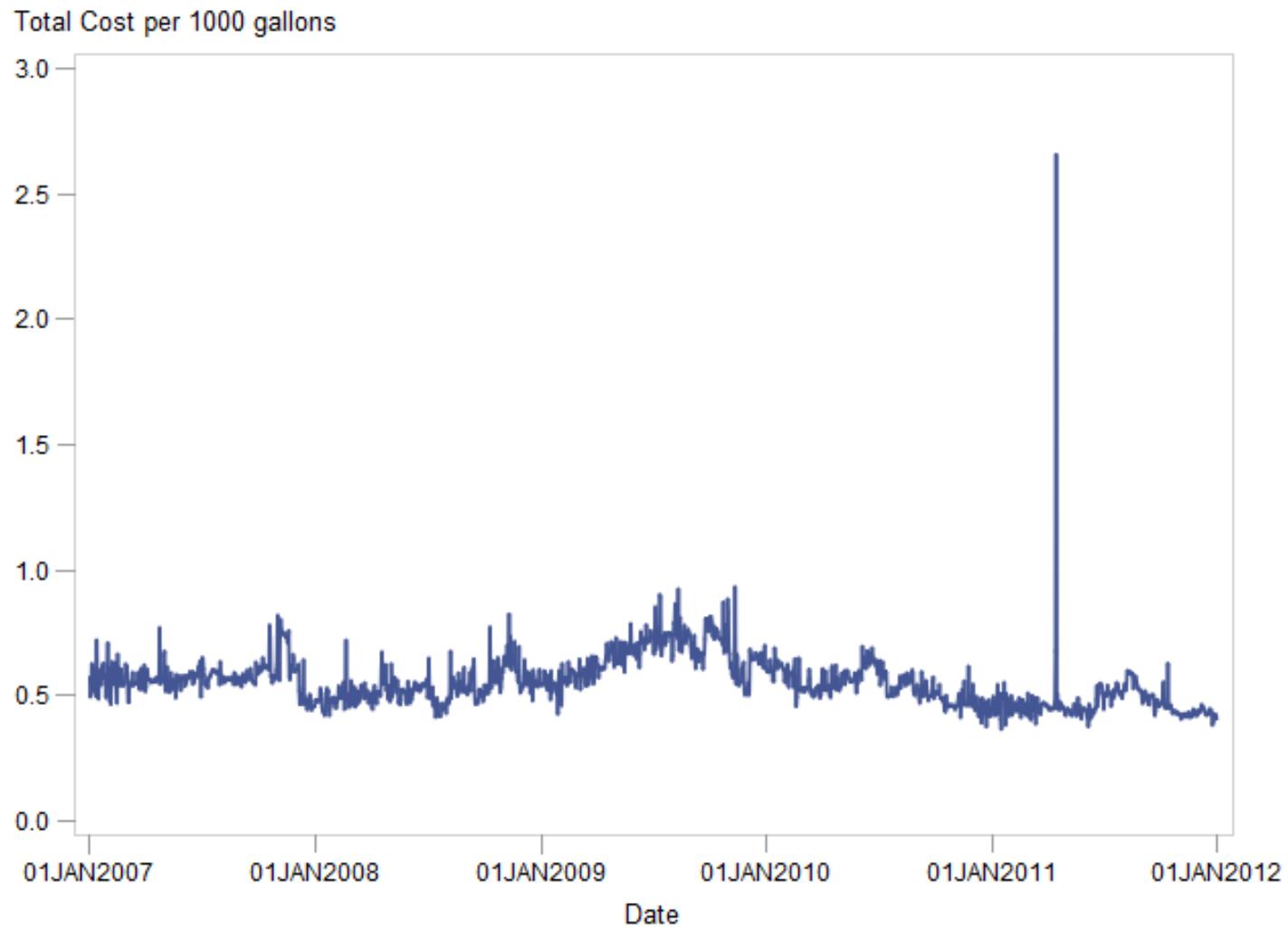
$$\theta_i = \alpha_0 + \sum_{j=1}^d \alpha_j f_j(i)$$

# Time Series Model

- Analyze daily costs using Error Correction Model (ECM).
  - Several commonly used time series models ignore the long-run equilibrium relationships predicted by economic theory
  - Long-run and short-run effects
- Changes in one independent variable can have immediate effect on treatment costs, a long-run effect, or both.
- **Period 1**

$$\frac{Cost}{1000 \text{ gallons}} = f \left( \begin{array}{l} FINAL, RAWTOC, TURB, pH, \\ GCPOOL, \\ TEMP, ActualTOC, \\ SPRSUM, CY07 - CY10, PROCESS \end{array} \right)$$

# Total Cost per 1000 Gallons of Finished Flow (Aug 2012\$)



## ECM Results, Period 1

- Cost of treating turbid water depends on current turbidity and water previously treated
- 1% decrease in TURB( $\approx 0.11\text{NTU}$ ) leads to an immediate decrease of 0.02% in TC/1000 gal. with another 0.1% decrease over future time. Total effect is 0.11%
- 1% decrease in turbidity, leads to \$1120 annual decrease in treatment costs



# Link Water Quality to Load Reductions

- How does total phosphorus (TP) load impact turbidity in source (raw) water?
  - Chose TP load because its affinity for natural clay particles and its link to harmful algal blooms
  - Ideal: Link daily nutrient/sediment results to daily turbidity (TURB)
  - TP load is weekly measurement of daily grab sample
- **Period 1:** Time series analysis (polynomial distributed lag model)
  - $TURB=f(TPLOAD, GCPOOL, SPR, CY09, CY10)$
- 1% reduction in TPLOAD ( $\approx 10.74$  lbs/day) leads to 0.15% decrease in turbidity over the long-run (5 weeks)
  - \$168 in treatment cost savings from 1% annual reduction in TP load
  - Cost to control 10.74 lbs/day for a year using cover crops = \$11763--\$105867
  - Price/lb = \$3—\$27;  $10.74 * 365 * \text{price}$



# ECM results, Period 2

GARCH Estimates					
SSE	12.2817467	Observations	1093		
MSE	0.01124	Uncond Var	0.01109977		
Log Likelihood	1313.45428	Total R-Square	0.9213		
SBC	-2458.9882	AIC	-2578.9086		
MAE	0.05345213	AIACC	-2577.785		
MAPE	271.364121	HQC	-2533.5276		
		Normality Test	7880.6867		
		Pr > ChiSq	<.0001		

Variable	DF	Parameter Estimates			Approx Pr >  t
		Estimate	Standard Error	t Value	
Intercept	1	0.4533	0.1704	2.66	0.0078
l_finaldiff	1	-0.7690	0.0886	-8.68	<.0001
l_rawtocdiff	1	0.1348	0.0748	1.80	0.0716
turbdiff	1	-0.000989	0.000429	-2.31	0.0211
phdiff	1	0.002229	0.0189	0.12	0.9061
p_sdiff	1	0.001056	0.001985	0.53	0.5948
tocdumdiff	1	0.001898	0.006447	0.29	0.7684
AOsep2013diff	1	0.6069	0.2710	2.24	0.0251
AO2015_2diff	1	0.8774	0.1912	4.59	<.0001
ll_total1realdiff	1	-0.1886	0.0330	-5.71	<.0001
ll2_total1real	1	-0.0642	0.0206	-3.12	0.0018
ll_final	1	-0.0896	0.0262	-3.43	0.0006
ll_rawtoc	1	0.0362	0.0304	1.19	0.2340
lag_turb	1	-0.001033	0.000632	-1.64	0.1020
lag_ph	1	0.0249	0.009419	2.64	0.0083
lag_ps	1	0.000694	0.000921	0.75	0.4510
lag_qtr2	1	0.0156	0.006682	2.33	0.0197
lag_tocdum	1	0.0106	0.009790	1.09	0.2777
lag_cy13	1	0.009214	0.0132	0.70	0.4842
lag_cy14	1	-0.004058	0.004439	-0.91	0.3606
lag_AOsep2013	1	0.4137	0.2155	1.92	0.0549
lag_AO2015_2	1	1.3004	0.4946	2.63	0.0086
ARCH0	1	0.003265	0.000645	5.06	<.0001
ARCH1	1	0.7058	0.3127	2.26	0.0240



# PDL results, Period 2

Estimate of Lag Distribution					
Variable	Estimate	Standard	t Value	Approx	0.314E-7
		Error		Pr >  t	
tploadlb(0)	0.00000152	9.26E-06	0.16	0.8698	**
tploadlb(1)	0.000008634	8.14E-06	1.06	0.2898	*****
tploadlb(2)	0.000014784	8.74E-06	1.69	0.092	*****
tploadlb(3)	0.000019969	9.73E-06	2.05	0.0413	*****
tploadlb(4)	0.00002419	0.0000103	2.34	0.0203	*****
tploadlb(5)	0.000027446	0.0000103	2.66	0.0084	*****
tploadlb(6)	0.000029738	9.69E-06	3.07	0.0024	*****
tploadlb(7)	0.000031066	8.73E-06	3.56	0.0004	*****
tploadlb(8)	0.000031429	8.29E-06	3.79	0.0002	*****
tploadlb(9)	0.000030827	9.68E-06	3.18	0.0016	*****
Convert to elasticity					
tploadlb(0)	0.00000152	0			
tploadlb(1)	0.000008634	0			
tploadlb(2)	0.000014784	0.002650673		Cost MGD	Cost day
tploadlb(3)	0.000019969	0.003580309		810	2745.9
tploadlb(4)	0.00002419	0.004337107		1002254	1351.038
tploadlb(5)	0.000027446	0.004920886			29270.95 L-R pH cost
tploadlb(6)	0.000029738	0.005331826			1%TPLOAD
tploadlb(7)	0.000031066	0.005569928			0.507353
tploadlb(8)	0.000031429	0.005635011			
tploadlb(9)	0.000030827	0.005527077			
		0.037552817			