

Motivation

- . Reservoir recreation is a highly-valued ecosystem s
- . However, visits to Oklahoma state parks adjacent to voirs has fallen by half since 1999.
- . Decreased visitation could be due to environmental

e Impact of Climate Varia the Demand for R

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This poster presents research on the impact of tion in Oklahoma. Increased air temperatures clining lake water conditions and algae bloom States. We relate state park visitation at reserving events. Our visitation model is estimated

tion and Water Condition eservoir Recreation

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ostract

f climate and water conditions on reservoir recreaare known to impact recreation demand while den events are a growing concern in the United voirs to water levels, water quality and swim warnusing panel data on Oklahoma park visits with a



Summary of the Results

We find the decline in visitation parks is not entirely celated to climate and water conditions.

Nevertheless, there is strong evidence air temperature Influences visitation, even controlling for seasonality

changes and thus declines in ecosystem service value



les.	time series from 1998 to 2015.			
400000 350000 250000 200000 150000 100000 50000	Lakes Broken Bow Eufaula Foss Fort Cobb Fort Gibson Grand Lake Hudson Keystone Spavinaw Tenkiller Tenkiller Texoma Thunderbird Tom Steed	State Park		
		Visitation Model S	D	

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ecification and Results

- (e.g. the popularity of summer months for outdoor recreation).
- There is modest evidence changes in water elevation levels relative to normal levels affects visitation.
- There is mixed evidence that BGA events and turbidity affect visitation.
- There is no evidence that the levels of rainfall and wind speed events over a month affect visitation.

viscussion and Conclusions (Preliminary)

Lake levels affect visitation. A 1-foot shift away from

- . Lake levels, blue-green algae (BGA) blooms and w clarity (turbidity) influence park visitation.
- . Measures of climate variability including temperature rain and wind-speed influence park visitation.
- . Changes in above factors contribute to visitation de

Data and Methods

Oklahoma Tourism and Recreation Department pro monthly visit data for 20 parks on 13 lakes, from '9
The Oklahoma Mesonet provided temperature, rain wind data.

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vided 8-'15. and . We estimate a log-linear visitation model usi ation as an exponential function of climate as $ln(Visits_{ij}) = \beta_1 |\Delta Lakelevel_{ij}| + \beta_2 Turbidity_{ij}$ $\beta_4 Temp_{ij} + \beta_5 Temp_{ij}^2 + \beta_6 Rain_{ij} + \beta_7$

. 4 versions of model: with and without the tin

	Model 1—No time trend		Model 2—Tin trend		
Variable	Parameter	P-value	Parameter	P-v	
Difference in lake level from normal	-0.019	0.021	-0.012	С	
BGA event	-0.429	0.000	-0.057	С	
Turbidity					

- ng OLS. The results are similar to modeling visitnd water conditions, estimated using NLS.
- $+\beta_3 BGAevent_{ij}+$
- Wind_{ij}+ β_8 Timetrend_i+ $\sum_i \Phi(Month_i)$ + $\sum_j \Phi(Park_j)$
- ne trend, and with and without turbidity (MI):

ne	Model 3—No time trend, fill in missing turbidity with MI		Model 4—Time trend, fill in missing turbidity with MI	
alue	Parameter	P-value	Parameter	P-value
.101	-0.021	0.017	-0.012	0.101
.625	-0.531	0.000	-0.067	0.576
	-0.005	0.080	-0.001	0.674

- normal elevation results in a 1%-2% decline in visits.
- Air temperatures affect visitation but the effect is noninear. Visits increase when temps rise from a low base, but decrease when temps rise from a high base.
- The significance of the time trend suggests part of the visitation decline is due to factors not accounted for as variables in the model.
- The occurrence of BGA events has increased over time, but the model does not provide strong evidence BGA events are impacting visitation.

Future Directions

- . Army Corps of Engineers provided water level and bloom data.
- Turbidity data came primarily from Oklahoma Wate source Board. <u>However, there was a large fraction of months and parks with no data</u>, so multiple imputat (MI) (Rubin, 1987) was used to fill in for missing d

<u>References</u>

. D.B. Rubin. *Multiple Imputation for Nonresponse in Surveys*. John Wil Sons, New York, 1987.

The participants in this project would like to acknowledge that possibility was made through funding provided by Oklahoma NSF EPSCOR award number IIA-1301789

BGA	Avg rainfall	-0.110	0.665	-0.238	0
	Avg wind speed	-0.043	0.022	0.012	0
er Re-	Avg air temp	0.097	0.001	0.021	0
nf	Avg air temp ²	-0.001	0.005	-0.0005	0
<u>)1</u>	Time trend			-0.006	0
1011	Observations		3,407		3
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. A parameter with a negative estimate provides evidence that provides evidence that variable has a positive effect on visita . All p-values are calculated under the null hypothesis that the

The time trend variable captures the effect of a constant annu itation unrelated to the other variables in the model.

.374	-0.030	0.561	-0.280	0.316
.963	-0.030	0.103	0.002	0.891
.002	0.093	0.003	0.070	0.009
.013	-0.001	0.016	-0.0004	0.036
.000			-0.006	0.000
,407		3,265		3,265

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variable has a negative effect on visitation, while a positive estimate tion.

true value is zero.

al change (possibly related to unobserved influential factors) in vis-

A lack of turbidity data for many park-month combinations motivated use of MI. We are working with other EPSCoR researchers to instead fill in this missng turbidity data with predictions from LandSat imngery.

The results of this study will be presented at the Southern Agricultural Economics Association 2016 Annual Meeting in San Antonio.