Environmental Determinants of Lake Trophic Status in the Con-

terminous United States: A Data Mining Approach

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1 Abstract

- Keywords: National Lakes Assessment, Cyanobacteria, Chlorophyl a, National Land Cover
- 13 Dataset, Random Forest, Data Mining

14 Introduction

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- Productivity in lentic systems is often categorized across a range of tropic states (e.g. the tropic continuum) from early successional (i.e. oligotrophic)to late successional lakes (i.e. hypereutrophic) (Carlson 1977). Lakes naturally occur across the range of trophic state and higher primary productiv-17 ity is not necessarily a predictor of poor ecological condition. Lakes that are naturally oligotrophic occur in nutrient poor areas or have a more recent geologic history. These lakes are often found in higher elevations, have clear water, and are often favored for drinking water or direct contact recreation (e.g. swimming). Lakes with higher productivity (e.g. eutrophic lakes) have greater 21 nutrient loads, tend to be less clear, have greater density of aquatic plants, and often support more diverse and abundant fish communities. Lakes will naturally shift to higher trophic states but this 23 is a slow process. Given this fact, monitoring trophic state allows the identification of rapid shifts in trophic state or locating lakes with unusually high productivity (e.g. hypereutrophic). These cases are indicative of lakes under greater anthropogenic nutrient loads, also known as cultural eutrophication, and are more likely to be at risk of fish kills, fouling, and harmful algal blooms[Smith 27 (1998); smith 1999 eutrophication; smith 2006 eutrophication].
- Given the association between trophic state and many ecosystem services and disservices, being able to model trophic state Most studies of trophic state are limited in spatial extent and don't look for broad scale patterns of variables that drive trophic state Most studies of trophic state focus on in-lake variables (i.e. nurients), limited ability to predict over large regions
 - We take advanatage of one the first complete national scale efforts monitoring lakes to try
 and discern broad patterns in both in-lake parameters that drive trophic state and landscape
 level parameters that might also drive trophic state

- Our primary question is, at the national scale, what are the primary determinants of lake trophic status?
 - Can those determinants be used to predict trophic state with an acceptable level of accuracy?
- Determinants include, chemical and physical parameters of the lake water column and land use/land cover. Lake trophic status defined by Chl a.

41 Methods

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Data and Study Area

- The two primary sources of data for this study are the National Lakes Assessment (NLA) data and
- the National Land Cover Dataset (NLCD) (USEPA 2009). Both datasets are national in scale and
- provide a unique snapshot view of the condition of United States' lakes and the patterns of the lakes
- surrounding landscape.
- 47 The NLA data were collected during the summer of 2007 and the final data were released in
- ⁴⁸ 2009. With consistent methods and metrics collected at 1056 locations across the conterminous
- 49 United States, the NLA provides a unique opportunity to examine continental scale patterns in lake
- productivity. The NLA collected data on biophysical meausers of lake water quality and habitat. For
- this analysis we primarily examined the water quality measurements from the NLA [TABLE REF].
- Adding to the monitoring data collected via the NLA, we use the 2006 NLCD data to examine the
- possible landscape-level drivers of trophic status in lakes. The NLCD is a nationally collected land
- use land cover dataset that also provides estimates of impervious surface. We collected total land
- use land cover and total percent impervious surface within the surroundin landscape of the lake.
- We defined the surrounding landscape of a lake with three different buffer distances: 300 meters,
- 1500 meters, and 2500 meters. The various distances were used to tease out differences in local
- ⁵⁸ landscape effects versus larger landscape-level effects.

59 Defining Trophic State

- 60 The dependent variable for this effort is lake trophic state. Trophic state is usually defined over four
- 61 levels: oligotrphic, mesotrophic, eutorphic, and hypereutrophic. Commonly, cut-off values for each
- of these four levels may be specified with nitrogen concnetration, phosphorus concentration, secchi
- depth, or chlorphyll a concentration (Carlson 1977; USEPA 2009). As this study is based largely
- from the NLA we use the NLA definition of trophic state based on the chlorophyll a concentrations
- 65 (Table).

Trophic State	Cut-off
oligotrophic	<= 0.2
mesotrophic	>2-7

Trophic State	Cut-off
eutrophic	>7-30
hypereutrophic	>30

66 Variable Selection

A strength of random forest is its ability to handle numerous correlated variables without a decrease 67 in prediction accuracy. Yet the number of redundant correlated predictor variables in our data requires a cursory reduction through the described variable selection method. To do this we 69 examine the correlation between log transformed chlorophyll a concentration and each of the log 70 transformed variables. The rationale behind this selection method is to discard variables with little 71 to no association with chlorphyll a and thus trophic state. Variables that explained less than 5% 72 of the variance (i.e. a pairwise correlation of less than 0.22) were assumed to not be associated 73 with cholorophyll a concentration and were removed from further consideration. Additionally, 74 variables measuring different attributes of the same distribution (e.g. minimum, maximum or mean temperature) were selected based on the variable with the strongest corelation with chlorophyll a. Lastly, the remaining predictor variables that are highly correlated with one another should not be included in the initial set of variables passed to the random forest, unless sepcified by domain 78 knowledge. As such we examine the pairwise correlations of these remaining variables and make a determination, as determined by knowledge of the system, as to which variables to retain.

81 Random Forest

As stated above, our goal is to explore relative variable importance in determination of lake trophic status. We selected random forest as the statistical analysis approach since random forest provides a robust measure of variable importance. Random forest is a machine learning algorithm that aggregates numerous decision trees in order to obtain to a consensus prediction of response categories. For every tree in the forest, the sample data are bootstrapped and subsets of predictor variables are randomly selected. Tree bifurcations are based on the best splits given the random subset of predictor variables and grown without pruning.

89 Variable Importance

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- How to use for variable selection
- what we used to identify important variables

2 Predicted Trophic State

- How random forests makes final predictions,
 - what we used to assess accuracy, etc.

95 Results

96 Summary Statistics

- Narrative summary.
- Table

99 Variable Selection

- Which variables were selected to include, and why, in the Random Forest.
- Table.

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• Pairs plot of selected variables showing little/weak association between selected variables.

104 Random Forest

- Summary of Random Forest model (number of Params, total oob, etc.)
- 106 Variable Importance
- Narrative description of variables.
- Table of Variables with gini or percent explained.
- 109 Predicted Trophic State
- Summary stats of percent of lakes in each class
- Confusion matrix of predicted with actual.

Discussion

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- What worked
- What didnt
 - What are the determinants and why improtant
 - How can this be expanded to other non-monitored lakes?
- What else can Trophic State tell us?
- Cyanobacteria association with?
- CDF Plots

Acknowledgements

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