

Prediction of potential areas of sinkhole development in southwestern Indiana using multiple regression analysis

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Abstract

Results of a multiple regression analysis, including prediction intervals, were used to develop a probability-based map of potential areas of sinkhole development in southwestern Indiana based on landscape-level variables as predictors. The dependent variable used in the analysis was the log of the sinkhole density (number of sinkholes/km²), derived from an inventory of 154,925 sinkholes. The analysis was conducted for five large (8-digit scale Hydrologic Unit Code) watersheds in southwestern Indiana and northern Kentucky.

The study area was divided into regular units of analysis at a scale of a quarter section in the Public Land Survey System (160 acres or 0.65 km²). For this analysis, 28,832 analysis cells were used. The conceptual model considers karst development to be more probable near specific geologic materials or features, especially in humid-temperate climates. The final best-fit equation accounted for 67% of the observed variability in the log sinkhole density and contained 14 statistically significant (99% confidence level) independent variables; five were related to lithology (including one interaction term, which combined geology and soils), two were land-use classes, and seven were related to terrain. Influential variables in the conceptual model were limestone, soil parent materials derived from limestone dissolution, and soil moisture and topographic parameters that might describe preferential flow to or from sinkhole features. The spatial distribution of potential sinkhole-development risk was mapped using 99% confidence intervals for the predicted log sinkhole density values.

The resultant distribution of sinkhole-development risk exists within a well-defined karst zone that trends from northwest to southeast along the Mitchell Plateau. The "extremely high" risk category has a high density of existing karst features, covering 2,617 km² (14%) of the study area. The "high" risk category borders the extremely high category, covering an area of 2,471 km² (13%). The "moderate" risk category (5,027 km², 27%) includes sinkholes developed in discontinuous limestone units as well as near faults where offsets have brought limestone closer to the surface. The "low" risk category covers 46% of the study area (8,554 km², 46%), and is dominated by noncarbonate bedrock geology, thick soil, or loess cover.

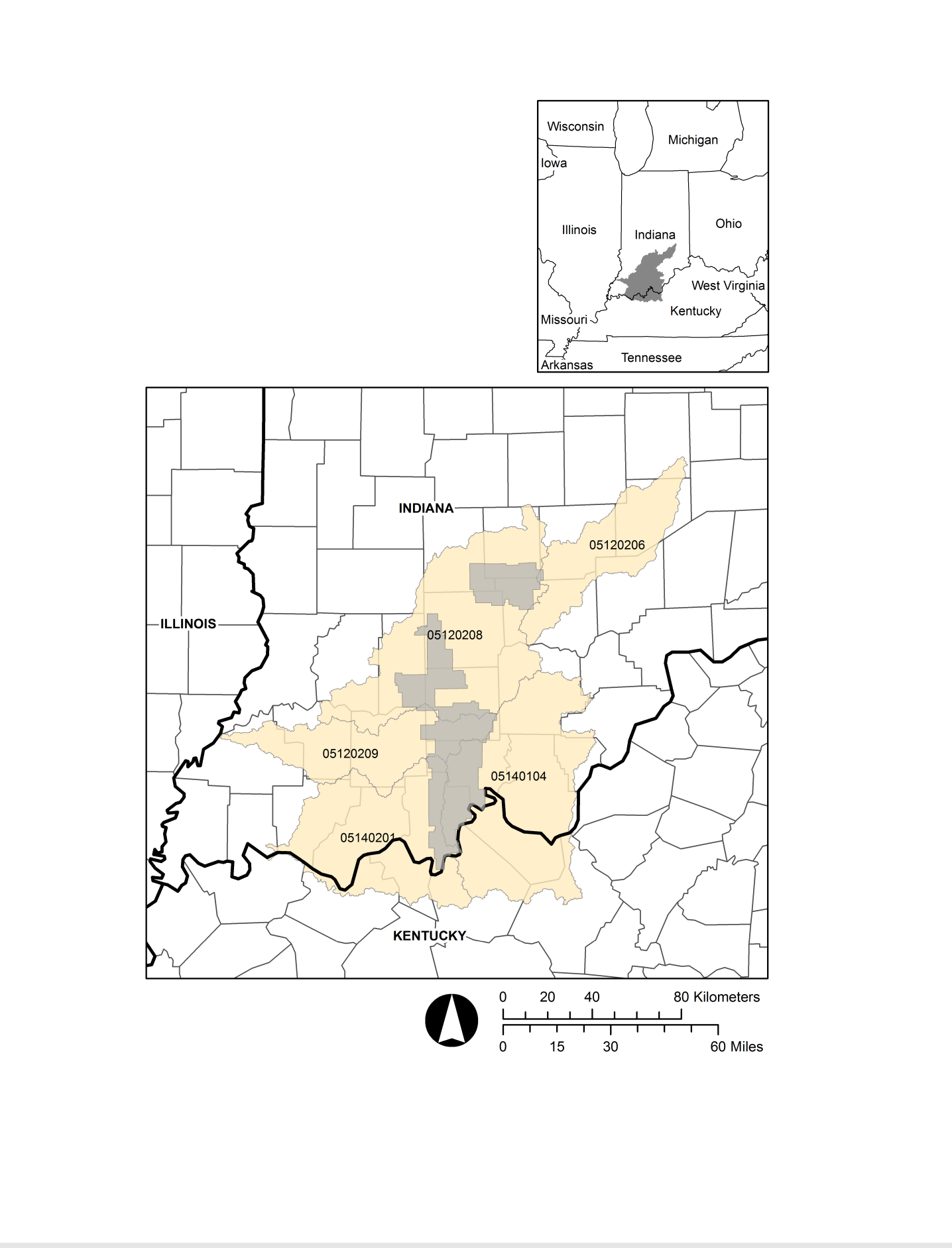


Figure 1. 8-digit watershed boundaries (HUC 8) encompassing the Hoosier National Forest (HNF) in southwestern Indiana. Watersheds are shown in yellow; HNF boundary shown in gray.

Conceptual Model

The regression analysis is based on a conceptual model that interprets karst development to be more probable in or near specific geologic materials or features, especially in humid-temperate climates, such as that of the study area. The independent variables that were hypothesized to most strongly influence sinkhole development include: limestone bedrock geology, soil parent materials that were derived from limestone dissolution, and soil moisture and topographic parameters that might describe preferential flow to or from sinkhole features. A small subset of sinkholes forms along fault lines in Indiana, so an effort was made to capture the variability related to structural offsets. The null hypothesis is that sinkholes are not related to landscape geologic or land-cover variables. The dependent variable in the study was the log of sinkhole density, which was determined after conducting an inventory of mapped and modeled sinkholes in southern Indiana and northern Kentucky in 2010 and 2011. The resultant database contained 154,925 sinkholes from which sinkhole density was determined.

The number of measurable contributing variables that can be captured in spatial (geographic information system) formats is substantial. For this study, the attributes that were considered are shown in Table 1, accompanied by source and scale information. For each variable evaluated, depending on the source scale of the original data, GIS data layers were created in vector or 30-meter raster formats. The data were aggregated by the regularized analysis cells either by conducting a spatial join to variables in vector formats, or by conducting a zonal statistical analysis to derive a mean value for variables in raster formats. For this analysis, 28,832 analysis cells were used.

Type	Variable	Source	Scale	Units	
Physical Characteristics	Catchment area	National hydrography dataset (NHD)	1 arc-second (30-m)	Sq kilometers	*
	Elevation	National Elevation Dataset (NED)	1 arc-second (30-m)	Meters	*
	Slope	NED derivative	1 arc-second (30-m)	Gradient (dimensionless)	*
	Aspect	NED derivative	1 arc-second (30-m)	degrees	*
	Curvature	NED derivative	1 arc-second (30-m)	dimensionless	*
	Compound Topographic Index (CTI)	NED derivative	1 arc-second (30-m)	dimensionless	*
	Direct Insolation duration	NED derivative	1 arc-second (30-m)	Watt hours per square meter (WH/m ²)	*
	Flow accumulation	NED derivative	1 arc-second (30-m)	Count	*
	Local contributing area	NED derivative	1 arc-second (30-m)	Sq meters	*
	Floodplain proximity	Federal Emergency Management Agency (FEMA)	500K	Meters	*
Hydrologic	Hydrologic variables (modeled)	United States Geological Survey (USGS) SPARROW model	1 arc-second (30-m)	mm/year % in/year	
	Geology	Indiana Geological Survey (IGS) and Kentucky Geological Survey (KGS)	250K (Indiana); 500K (Kentucky)	---	*
	Karst Proximity	IGS/KGS geology derivative	250K (Indiana); 500K (Kentucky)	Meters	*
	Carbonate	IGS/KGS geology derivative	250K (Indiana); 500K (Kentucky)	Percent	*
	Depth to limestone	IGS/KGS geology and NED derivative	250K (Indiana); 500K (Kentucky)	Meters	*
Geologic	Fault proximity	United States Geological Survey (USGS)	12K to 500K	Meters	*
	Geology and Soil *	IGS and NRCS derivative	12K – 500K	---	*
	Soil Parent Material	Natural Resources Conservation Service (NRCS) SSURGO	12K – 20K	---	*
	Available water storage 150cm	Natural Resources Conservation Service (NRCS) SSURGO	12K – 20K	---	*
	Soil proximity	NRCS derivative	12K – 20K	Meters	*
Land Cover	Land cover (15 classes evaluated)	National Land Cover Dataset (NLCD)	24K (30-m)	Area, sq meters	* Forest * Hay

Table 1. Variables considered. Only variables that demonstrated statistical significance (*) were used in the final analysis.

Statistical Model Background

The spatial data were prepared for input into a multivariate model, intended to investigate the predictive power of the landscape variables on sinkhole density, and therefore prediction of sinkhole-development risk. A multivariate method is appropriate for this type of analysis because errors (in this case, omissions) in the measurement of the dependent variable, sinkhole density, are probable, and the multiple regression is appropriate because random error computation is incorporated into the parameter estimates and significance tests.

The statistical regression model had the following form:

$$\log Y = b_0 + b_1 x_1 + b_2 x_2 + b_3 x_3 + b_4 x_4 + \dots b_m x_m + e$$

Where Y is sinkhole density, $x_1 - x_m$ are independent variables that are used to improve the prediction of sinkhole density and $b_0 - b_m$ are best-fit regression coefficients estimated using ordinary least squares, and e is an error term. The dependent variable was log-transformed because the range of observed density values covered several orders of magnitude, and the transformation provided a better model fit.

Multiple Regression Analysis

Each of the fourteen landscape variables was included in the regression analysis as an independent variable. The magnitude of the student's *t* statistic indicates the relative importance of each independent variable to the prediction of sinkhole density. The *t* values (Table 2) indicate that all of the variables were contributing significantly to the prediction at the 99% confidence level. This is not surprising given the unusually large sample size of 28,832 and the dependence of standard error on sample size. Figure 2 shows the independent variables in order of significance. The best-fit model combined variables that elucidated details from the geologic and soils data, with lesser influence from topographic and land-use variables. Although some variables have an equivalent effect on sinkhole development (e.g., soil parent material and proximity to important soil parent materials), the correlation matrix does not show interdependence, and the predictive power of the regression equation is diminished when either variable is removed. Land-use and land-cover classes were included to identify any relationships to the log of sinkhole density; of 15 land-cover variables, only mixed forest classes and hay were significant. Two variables intended to capture sinkhole development processes around faults, proximity to mapped faults and depth to limestone, were not statistically significant.

Variable ID	Variable name	Parameter Estimate, b	Standard Error (SE)	t-ratio (t=b/SE)
V ₁	Catchment area	5.93E-03	1.84E-04	32.15
V ₂	Elevation	-5.02E-03	2.06E-04	-24.38
V ₃	Aspect	-2.04E-03	2.01E-04	-10.12
V ₄	Curvature	3.13	4.41E-01	7.08
V ₅	CTI	9.26E-02	4.23E-03	21.89
V ₆	Insolation	1.60E-03	8.82E-05	18.09
V ₇	Carbonate	1.62E-02	3.35E-04	48.45
V ₈	Geology (Karst)	2.80E-01	3.71E-02	7.56
V ₉	Karst proximity	6.58E-05	1.56E-06	42.09
V ₁₀	Soil parent	4.05E-01	2.50E-02	16.21
V ₁₁	Geology & Soil	1.21E-01	2.56E-02	4.71
V ₁₂	Soil proximity	-2.28E-05	4.81E-07	-47.39
V ₁₃	Forest	6.16E-07	5.04E-08	12.23
V ₁₄	Hay	6.61E-07	5.24E-08	12.60

Table 2. Parameter estimates, standard errors, and t-ratios. The null hypothesis that the regression parameter equals zero is rejected at the 99% confidence level when the calculated t-ratio exceeds the critical value of 2.58 for degrees of freedom > 1000. Total sample size, N = 28,832. R² = 0.67. Multiple correlation coefficient, R = 0.82.

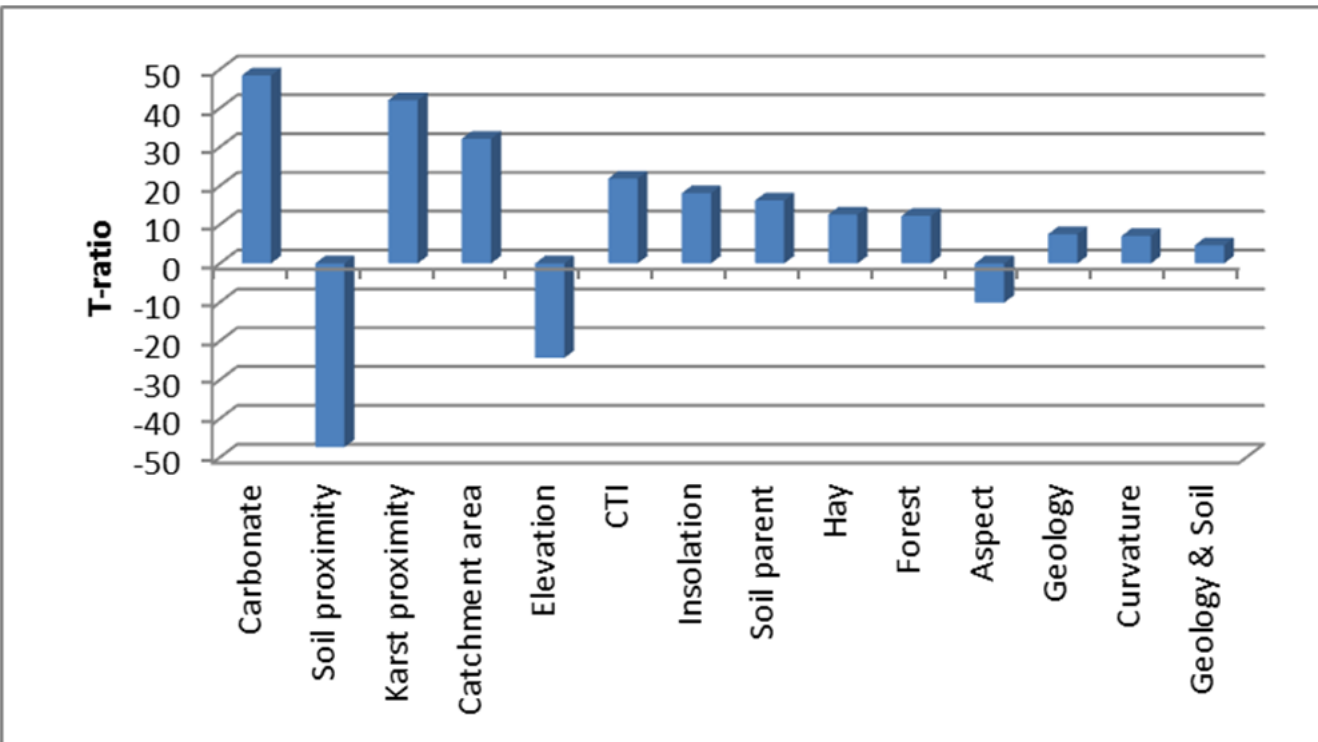


Figure 2. Computed t-ratios of significant parameters used in the multiple regression analysis.

