

# Andrew O Finley

## List of Publications by Year in descending order

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Version: 2024-02-01

81  
papers

3,901  
citations

147801

31  
h-index

133252

59  
g-index

83  
all docs

83  
docs citations

83  
times ranked

4157  
citing authors

#	ARTICLE	IF	CITATIONS
1	Gaussian Predictive Process Models for Large Spatial Data Sets. <i>Journal of the Royal Statistical Society Series B: Statistical Methodology</i> , 2008, 70, 825-848.	2.2	673
2	Hierarchical Nearest-Neighbor Gaussian Process Models for Large Geostatistical Datasets. <i>Journal of the American Statistical Association</i> , 2016, 111, 800-812.	3.1	335
3	A Case Study Competition Among Methods for Analyzing Large Spatial Data. <i>Journal of Agricultural, Biological, and Environmental Statistics</i> , 2019, 24, 398-425.	1.4	216
4	Improving the performance of predictive process modeling for large datasets. <i>Computational Statistics and Data Analysis</i> , 2009, 53, 2873-2884.	1.2	168
5	<code>spBayes</code> : An R Package for Univariate and Multivariate Hierarchical Point-referenced Spatial Models. <i>Journal of Statistical Software</i> , 2007, 19, 1-24.	3.7	158
6	Tropical tree growth is correlated with soil phosphorus, potassium, and calcium, though not for legumes. <i>Ecological Monographs</i> , 2012, 82, 189-203.	5.4	128
7	Comparing spatially-varying coefficients models for analysis of ecological data with non-stationary and anisotropic residual dependence. <i>Methods in Ecology and Evolution</i> , 2011, 2, 143-154.	5.2	125
8	Towards connecting biodiversity and geodiversity across scales with satellite remote sensing. <i>Global Ecology and Biogeography</i> , 2019, 28, 548-556.	5.8	87
9	<code>spBayes</code> for Large Univariate and Multivariate Point-Referenced Spatio-Temporal Data Models. <i>Journal of Statistical Software</i> , 2015, 63, .	3.7	85
10	rFIA: An R package for estimation of forest attributes with the US Forest Inventory and Analysis database. <i>Environmental Modelling and Software</i> , 2020, 127, 104664.	4.5	81
11	Addressing data integration challenges to link ecological processes across scales. <i>Frontiers in Ecology and the Environment</i> , 2021, 19, 30-38.	4.0	74
12	Nonseparable dynamic nearest neighbor Gaussian process models for large spatio-temporal data with an application to particulate matter analysis. <i>Annals of Applied Statistics</i> , 2016, 10, 1286-1316.	1.1	73
13	Predicting tree biomass growth in the temperate-boreal ecotone: Is tree size, age, competition, or climate response most important?. <i>Global Change Biology</i> , 2016, 22, 2138-2151.	9.5	71
14	Efficient Algorithms for Bayesian Nearest Neighbor Gaussian Processes. <i>Journal of Computational and Graphical Statistics</i> , 2019, 28, 401-414.	1.7	71
15	Working across space and time: nonstationarity in ecological research and application. <i>Frontiers in Ecology and the Environment</i> , 2021, 19, 66-72.	4.0	69
16	Should species distribution models account for spatial autocorrelation? A test of model projections across eight millennia of climate change. <i>Global Ecology and Biogeography</i> , 2013, 22, 760-771.	5.8	67
17	An analysis of asthma hospitalizations, air pollution, and weather conditions in Los Angeles County, California. <i>Science of the Total Environment</i> , 2012, 425, 110-118.	8.0	65
18	Approaches to advance scientific understanding of macrosystems ecology. <i>Frontiers in Ecology and the Environment</i> , 2014, 12, 15-23.	4.0	57

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19	Boreal tree growth exhibits decadal-scale ecological memory to drought and insect defoliation, but no negative response to their interaction. <i>Journal of Ecology</i> , 2019, 107, 1288-1301.	4.0	49
20	Over half of western United States' most abundant tree species in decline. <i>Nature Communications</i> , 2021, 12, 451.	12.8	48
21	Approximate Bayesian inference for large spatial datasets using predictive process models. <i>Computational Statistics and Data Analysis</i> , 2012, 56, 1362-1380.	1.2	47
22	Hierarchical Spatial Process Models for Multiple Traits in Large Genetic Trials. <i>Journal of the American Statistical Association</i> , 2010, 105, 506-521.	3.1	44
23	On nearest-neighbor Gaussian process models for massive spatial data. <i>Wiley Interdisciplinary Reviews: Computational Statistics</i> , 2016, 8, 162-171.	3.9	44
24	LiDAR based prediction of forest biomass using hierarchical models with spatially varying coefficients. <i>Remote Sensing of Environment</i> , 2015, 169, 113-127.	11.0	40
25	Hierarchical spatial models for predicting tree species assemblages across large domains. <i>Annals of Applied Statistics</i> , 2009, 3, 1052-1079.	1.1	39
26	Bayesian dynamic modeling for large space-time datasets using Gaussian predictive processes. <i>Journal of Geographical Systems</i> , 2012, 14, 29-47.	3.1	39
27	Geostatistical estimation of forest biomass in interior Alaska combining Landsat-derived tree cover, sampled airborne lidar and field observations. <i>Remote Sensing of Environment</i> , 2018, 212, 212-230.	11.0	39
28	Adaptive Gaussian predictive process models for large spatial datasets. <i>Environmetrics</i> , 2011, 22, 997-1007.	1.4	37
29	Strategies for minimizing sample size for use in airborne LiDAR-based forest inventory. <i>Forest Ecology and Management</i> , 2013, 292, 75-85.	3.2	37
30	Modeling forest biomass and growth: Coupling long-term inventory and LiDAR data. <i>Remote Sensing of Environment</i> , 2016, 182, 1-12.	11.0	36
31	Bayesian multivariate process modeling for prediction of forest attributes. <i>Journal of Agricultural, Biological, and Environmental Statistics</i> , 2008, 13, 60-83.	1.4	34
32	Synergistic effects of climate and land cover: grassland birds are more vulnerable to climate change. <i>Landscape Ecology</i> , 2016, 31, 2275-2290.	4.2	33
33	spOccupancy: An R package for single-species, multi-species, and integrated spatial occupancy models. <i>Methods in Ecology and Evolution</i> , 2022, 13, 1670-1678.	5.2	32
34	Spatial regression methods capture prediction uncertainty in species distribution model projections through time. <i>Global Ecology and Biogeography</i> , 2013, 22, 242-251.	5.8	29
35	Landscape fragmentation affects responses of avian communities to climate change. <i>Global Change Biology</i> , 2015, 21, 2942-2953.	9.5	29
36	Beyond counts and averages: Relating geodiversity to dimensions of biodiversity. <i>Global Ecology and Biogeography</i> , 2020, 29, 696-710.	5.8	29

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37	Ecological forecasting of tree growth: Regional fusion of treeâ€ring and forest inventory data to quantify drivers and characterize uncertainty. <i>Global Change Biology</i> , 2022, 28, 2442-2460.	9.5	29
38	A Bayesian approach to multi-source forest area estimation. <i>Environmental and Ecological Statistics</i> , 2008, 15, 241-258.	3.5	28
39	Nonlinear hierarchical models for predicting cover crop biomass using Normalized Difference Vegetation Index. <i>Remote Sensing of Environment</i> , 2010, 114, 2833-2840.	11.0	28
40	A Hierarchical Model for Quantifying Forest Variables Over Large Heterogeneous Landscapes With Uncertain Forest Areas. <i>Journal of the American Statistical Association</i> , 2011, 106, 31-48.	3.1	28
41	Variable effects of climate on forest growth in relation to climate extremes, disturbance, and forest dynamics. <i>Ecological Applications</i> , 2017, 27, 1082-1095.	3.8	27
42	Integrating forest inventory and analysis data into a LIDAR-based carbon monitoring system. <i>Carbon Balance and Management</i> , 2014, 9, 3.	3.2	26
43	Seedling survival responses to conspecific density, soil nutrients, and irradiance vary with age in a tropical forest. <i>Ecology</i> , 2016, 97, 2406-2415.	3.2	25
44	Multivariate Spatial Regression Models for Predicting Individual Tree Structure Variables Using LiDAR Data. <i>IEEE Journal of Selected Topics in Applied Earth Observations and Remote Sensing</i> , 2013, 6, 6-14.	4.9	23
45	Linear Models for Airborne-Laser-Scanning-Based Operational Forest Inventory With Small Field Sample Size and Highly Correlated LiDAR Data. <i>IEEE Transactions on Geoscience and Remote Sensing</i> , 2015, 53, 5600-5612.	6.3	23
46	Hierarchical Spatial Modeling of Additive and Dominance Genetic Variance for Large Spatial Trial Datasets. <i>Biometrics</i> , 2009, 65, 441-451.	1.4	20
47	Comparing and Blending Regional Climate Model Predictions for the American Southwest. <i>Journal of Agricultural, Biological, and Environmental Statistics</i> , 2011, 16, 586-605.	1.4	19
48	Highly Scalable Bayesian Geostatistical Modeling via Meshed Gaussian Processes on Partitioned Domains. <i>Journal of the American Statistical Association</i> , 2022, 117, 969-982.	3.1	19
49	Variational Bayesian methods for spatial data analysis. <i>Computational Statistics and Data Analysis</i> , 2011, 55, 3197-3217.	1.2	18
50	Hierarchical Bayesian spatial models for predicting multiple forest variables using waveform LiDAR, hyperspectral imagery, and large inventory datasets. <i>International Journal of Applied Earth Observation and Geoinformation</i> , 2013, 22, 147-160.	2.8	18
51	Spatial Variation in Nutrient and Water Color Effects on Lake Chlorophyll at Macroscales. <i>PLoS ONE</i> , 2016, 11, e0164592.	2.5	18
52	Bayesian spatially varying coefficient models in the spBayes R package. <i>Environmental Modelling and Software</i> , 2020, 125, 104608.	4.5	18
53	Bayesian hierarchical models for spatially misaligned data in R. <i>Methods in Ecology and Evolution</i> , 2014, 5, 514-523.	5.2	16
54	Assessing soundscape disturbance through hierarchical models and acoustic indices: A case study on a shelterwood logged northern Michigan forest. <i>Ecological Indicators</i> , 2020, 113, 106244.	6.3	15

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55	Modeling Complex Spatial Dependencies: Low-Rank Spatially Varying Cross-Covariances With Application to Soil Nutrient Data. <i>Journal of Agricultural, Biological, and Environmental Statistics</i> , 2013, 18, 274-298.	1.4	14
56	Integrating automated acoustic vocalization data and point count surveys for estimation of bird abundance. <i>Methods in Ecology and Evolution</i> , 2021, 12, 1040-1049.	5.2	14
57	Spatial Factor Models for High-Dimensional and Large Spatial Data: An Application in Forest Variable Mapping. <i>Statistica Sinica</i> , 2019, 29, 1155-1180.	0.3	13
58	Bayesian multi-resolution modeling for spatially replicated data sets with application to forest biomass data. <i>Journal of Statistical Planning and Inference</i> , 2007, 137, 3193-3205.	0.6	12
59	Complementary strengths of spatially explicit and multi-species distribution models. <i>Ecography</i> , 2020, 43, 456-466.	4.5	11
60	Accounting for the space-varying nature of the relationships between temporal community turnover and the environment. <i>Ecography</i> , 2014, 37, 1073-1083.	4.5	10
61	Spatial Analysis of Anthropogenic Landscape Disturbance and Buruli Ulcer Disease in Benin. <i>PLoS Neglected Tropical Diseases</i> , 2015, 9, e0004123.	3.0	10
62	Spatial scaling of temporal changes in avian communities. <i>Global Ecology and Biogeography</i> , 2015, 24, 1236-1248.	5.8	9
63	A model-based approach to wildland fire reconstruction using sediment charcoal records. <i>Environmetrics</i> , 2017, 28, e2450.	1.4	9
64	Regional-based mitigation to reduce wildlife-vehicle collisions. <i>Journal of Wildlife Management</i> , 2018, 82, 756-765.	1.8	9
65	High-dimensional multivariate geostatistics: A Bayesian matrix-normal approach. <i>Environmetrics</i> , 2021, 32, e2675.	1.4	9
66	Spatial Modelling of Car Ownership Data: A Case Study from the United Kingdom. <i>Applied Spatial Analysis and Policy</i> , 2010, 3, 45-65.	2.0	8
67	Dynamic spatial regression models for space-varying forest stand tables. <i>Environmetrics</i> , 2014, 25, 596-609.	1.4	8
68	Joint hierarchical models for sparsely sampled high-dimensional LiDAR and forest variables. <i>Remote Sensing of Environment</i> , 2017, 190, 149-161.	11.0	8
69	Characterizing functional relationships between anthropogenic and biological sounds: a western New York state soundscape case study. <i>Landscape Ecology</i> , 2020, 35, 689-707.	4.2	8
70	Improving Crop Model Inference Through Bayesian Melding With Spatially Varying Parameters. <i>Journal of Agricultural, Biological, and Environmental Statistics</i> , 2011, 16, 453-474.	1.4	7
71	Assessing impact of exogenous features on biotic phenomena in the presence of strong spatial dependence: A lake sturgeon case study in natural stream settings. <i>PLoS ONE</i> , 2018, 13, e0204150.	2.5	6
72	Introduction to Bayesian Methods in Ecology and Natural Resources. , 2020, , .		6

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73	Trends in bird abundance differ among protected forests but not bird guilds. <i>Ecological Applications</i> , 2021, 31, e02377.	3.8	6
74	Hierarchical multiresolution approaches for dense point-level breast cancer treatment data. <i>Computational Statistics and Data Analysis</i> , 2008, 52, 2650-2668.	1.2	5
75	Observation-based blended projections from ensembles of regional climate models. <i>Climatic Change</i> , 2016, 138, 55-69.	3.6	5
76	Environmental controls on Landsat-derived phenoregions across an East African megatransect. <i>Ecosphere</i> , 2020, 11, e03143.	2.2	4
77	Remote Sensing of Geodiversity as a Link to Biodiversity. , 2020, , 225-253.		4
78	Estimating timber volume loss due to storm damage in Carinthia, Austria, using ALS/TLS and spatial regression models. <i>Forest Ecology and Management</i> , 2021, 502, 119714.	3.2	4
79	Simplifying Small Area Estimation With rFIA: A Demonstration of Tools and Techniques. <i>Frontiers in Forests and Global Change</i> , 2022, 5, .	2.3	3
80	Editors Are Editors, Not Oracles. <i>Bulletin of the Ecological Society of America</i> , 2014, 95, 342-346.	0.2	2
81	Spatial Linear Models. , 2020, , 155-174.		0