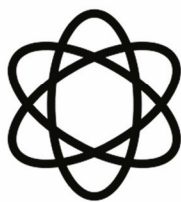




Data-Driven Earth Observation for Disaster Management

From Theory to Practical Applications



Edited by

Xiao Huang, Siqin Wang, Kleomenis Kalogeropoulos, Andreas Tsatsaris

Data-Driven Earth Observation for Disaster Management

Earth Observation

Data-Driven Earth Observation for Disaster Management

From Theory to Practical Applications

Edited by

Xiao Huang

Department of Environmental Sciences, Emory University, Atlanta, GA, United States

Siqin Wang

Spatial Sciences Institute, University of Southern California, Los Angeles, CA, United States

Kleomenis Kalogeropoulos

*Department of Surveying and Geoinformatics Engineering, University of West Attica,
Egaleo, Athens, Greece*

Andreas Tsatsaris

*Department of Surveying and Geoinformatics Engineering, University of West Attica,
Egaleo, Athens, Greece*

Series Editor

George P. Petropoulos



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Radarweg 29, PO Box 211, 1000 AE Amsterdam, Netherlands
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Contributors

- Mohamed Abdelkader** Department of Civil, Environmental and Ocean Engineering (CEOE), Stevens Institute of Technology, Hoboken, NJ; Department of Civil and Environmental Engineering, IIHR-Hydroscience & Engineering, The University of Iowa, Iowa City, IA, United States
- Aarnav Agrawal** Monta Vista High School, Cupertino, CA, United States
- Menna Farag Ahmed** Disaster Prevention Research Institute, Kyoto University, Uji-shi, Kyoto, Japan
- Nasrin Alamdari** Department of Civil and Environmental Engineering; Resilient Infrastructure and Disaster Response Center, Florida State University, Tallahassee, FL, United States
- Cesar Ivan Alvarez** Centre for Climate Resilience, University of Augsburg, Augsburg, Germany
- Yosuke Aoki** Earthquake Research Institute, University of Tokyo, Tokyo, Japan
- A. Asutosh** Aurassure Pvt. Ltd., Bhubaneswar, Odisha, India
- Mitra Nasr Azadani** Department of Civil and Environmental Engineering; Resilient Infrastructure and Disaster Response Center, Florida State University, Tallahassee, FL, United States
- Aradhana Basu** School of Sustainability, XIM University, Bhubaneswar, Odisha; Jindal School of Government and Public Policy, O.P. Jindal Global University, Sonapat, Haryana, India
- Filippo Maria Bianchi** Mathematics and Statistics, UiT The Arctic University of Norway; NORCE Norwegian Research Centre AS, Technology—Earth Observation Group, Tromsø, Norway
- Doan Van Binh** Faculty of Engineering, Vietnamese-German University, Ho Chi Minh City, Vietnam
- Guofeng Cao** Department of Geography, University of Colorado Boulder, Boulder, CO, United States
- Kanglin Chen** Department of Government, Harvard University, Cambridge, MA, United States
- Katerina Dermosinoglou** Department of Geography, Harokopio University of Athens, Athens, Greece
- Spyridon E. Detsikas** Department of Geography, Harokopio University of Athens, Athens, Greece
- Chunyu Diao** Department of Geography and Geographic Information Science, University of Illinois Urbana-Champaign, Urbana, IL, United States
- Giuseppe DiCaprio** PLANETGIS SKY S.R.L, Milan, Italy
- Qing Ding** College of Geo-Exploration Science and Technology, Jilin University, Changchun, China
- Nguyen Luyen Phuong Doan** Mekong Institute, Can Tho University, Can Tho City, Vietnam
- Chunyu Dong** School of Civil Engineering, Sun Yat-Sen University, Zhuhai, China
- Fang Fang** School of Public Policy and Urban Affairs, School of Criminology and Criminal Justice, Northeastern University, Boston, MA, United States
- Elissavet Feloni** Department of Surveying and Geoinformatics Engineering, University of West Attica, Athens, Greece

Contributors

Xiaokang Fu Center for Geographic Analysis, Harvard University, Cambridge, MA, United States; State Key Laboratory of Information Engineering in Surveying, Mapping and Remote Sensing, Wuhan University, Wuhan, China

Christian Geiß German Remote Sensing Data Center (DFD), German Aerospace Center (DLR), Oberpfaffenhofen; Department of Geography, University of Bonn, Bonn, Germany

Fatemeh Ghasempour Zonguldak Bulent Ecevit University, Zonguldak, Türkiye

Jakob Grahn NORCE Norwegian Research Centre AS, Technology—Earth Observation Group, Tromsø, Norway

Dayana Gualotuña Environmental Research Group on Sustainable Development (GIADES), Salesian Polytechnic University, Quito, Ecuador

Xulin Guo Department of Geography and Planning, University of Saskatchewan, Saskatoon, Canada

Thamanna Hari School of Sustainability, XIM University, Bhubaneswar, Odisha, India

Yaqian He Department of Geosciences, University of Arkansas, Fayetteville, GA; Department of Geosciences, University of Arkansas, Fayetteville, AR; Department of Geography, Indiana University Bloomington, Bloomington, IN, United States

Miaoqing Huang Department of Electrical Engineering and Computer Science, University of Arkansas, Fayetteville, AR, United States

Xiao Huang Department of Environmental Sciences, Emory University, Atlanta, GA, United States

Syed Usama Imtiaz Department of Civil and Environmental Engineering; Resilient Infrastructure and Disaster Response Center, Florida State University, Tallahassee, FL, United States

Kleomenis Kalogeropoulos Department of Surveying and Geoinformatics Engineering, University of West Attica, Egaleo, Athens, Greece

Sameh Ahmed Kantoush Disaster Prevention Research Institute, Kyoto University, Uji-shi, Kyoto, Japan

Masatoshi Katabuchi Yunnan Key Laboratory of Forest Ecosystem Stability and Global Change Response, Chinese Academy of Sciences, Mengla, Yunnan; University of Chinese Academy of Sciences, Beijing, China

Nitesh Khadka Institute of Mountain Hazards and Environment, Chinese Academy of Sciences, Chengdu; University of Chinese Academy of Sciences, Beijing, China

Steffen Knoblach GIScience Research Group, Institute of Geography, Heidelberg University; HeiGIT gGmbH at Heidelberg University, Heidelberg, Germany

Simran Koul Bren School of Environmental Science & Management, University of California, Santa Barbara, CA, United States

Senol Hakan Kutoglu Zonguldak Bulent Ecevit University, Zonguldak, Türkiye

Feng Lai School of Civil Engineering, Sun Yat-Sen University, Zhuhai, China

Feilin Lai Department of Earth and Space Sciences, Lamar University, Beaumont, TX, United States

Fa Li Department of Earth System Science, Stanford University, Stanford, CA, United States

Hao Li Department of Geography, National University of Singapore, Singapore

Jiwei Li School of Ocean, Futures, School of Earth and Space Exploration, Center for Global Discovery and Conservation Sciences, Arizona State University, Tempe, AZ, United States

Longxiang Li Gangarosa Department of Environmental Health, Emory University, Atlanta, GA, United States

- Shuai Li** Department of Geography, University of Florida, Gainesville, FL, United States; Yunnan Key Laboratory of Forest Ecosystem Stability and Global Change Response, Chinese Academy of Sciences, Mengla, Yunnan, China
- Zhiying Li** O'Neill School of Public and Environmental Affairs, Indiana University Bloomington, Bloomington, IN, United States
- Hao-Yu Liao** Agronomy Department, University of Florida, Gainesville, FL, United States
- Kairong Lin** School of Civil Engineering, Sun Yat-Sen University, Zhuhai, China
- Tao Liu** Warnell School of Forestry and Natural Resources, University of Georgia, Athens, GA, United States
- Xiao Liu** Georgia Institute of Technology, Industrial and Systems Engineering, Atlanta, GA, United States
- Santiago López** School of Interdisciplinary Arts and Sciences, University of Washington Bothell, Bothell, WA, United States
- Junyu Lu** School of Community Resources and Development, Arizona State University, Phoenix, AZ, United States
- Tianjun Lu** Epidemiology and Environmental Health, University of Kentucky, Lexington, KY, United States
- Ziqiang Ma** School of Earth and Space Sciences, Institute of Remote Sensing and Geographical Information Systems, Peking University, Beijing, China
- Grayson Russell Morgan** Department of Geography, Brigham Young University, Provo, UT, United States
- Rajasekhar Reddy Muttana** Jindal School of Government and Public Policy, O.P. Jindal Global University, Sonapat, Haryana, India
- Binh Quang Nguyen** Disaster Prevention Research Institute, Kyoto University, Uji-shi, Kyoto, Japan; The University of Danang – University of Science and Technology, Danang, Vietnam
- Patrick Pelizari** German Remote Sensing Data Center (DFD), German Aerospace Center (DLR), Oberpfaffenhofen, Bonn, Germany
- George P. Petropoulos** Department of Geography, Harokopio University of Athens, Athens, Greece
- Weikang Qian** Department of Geography, University of Florida, Gainesville, FL, United States
- Mengling Qiao** Department of Earth and Environmental Engineering, Columbia University, NY, United States
- Le Van Quyen** Faculty of Engineering, Vietnamese-German University, Ho Chi Minh City, Vietnam
- Mayra A. Román-Rivera** Geography and Sustainability, University of Tennessee, Knoxville, TN, United States
- Aliihsan Sekertekin** Department of Architecture and Town Planning, Iğdir University, Iğdir, Türkiye
- Shahrin Shahpar** Department of Electrical Engineering and Computer Science, University of Arkansas, Fayetteville, AR, United States
- Chao Song** Sichuan University, Chengdu, China
- Irini Soubry** Department of Geography and Planning, University of Saskatchewan, Saskatoon, Canada
- Xiaonan Tai** Department of Biological Sciences, New Jersey Institute of Technology, Newark, NJ, United States
- Maham Tariq** Department of Environmental Sciences, COMSATS University Islamabad, Vehari, Pakistan

Contributors

- Hannes Taubenböck** German Remote Sensing Data Center (DFD), German Aerospace Center (DLR), Oberpfaffenhofen, Bonn; Department of Remote Sensing, Institute of Geography and Geology, University of Würzburg, Würzburg, Germany
- Pratyush Tripathy** Department of Geography, University of California, Santa Barbara, CA, United States
- Andreas Tsatsaris** Department of Surveying and Geoinformatics Engineering, University of West Attica, Egaleo, Athens, Greece
- Luc Anh Tuan** Faculty of Engineering, Vietnamese-German University, Ho Chi Minh City, Vietnam
- Pawel Tysiac** Faculty of Civil and Environmental Engineering, EKO-TECH Centre, Gdansk University of Technology, Gdansk, Poland
- Mirza Waleed** Department of Geography, Hong Kong Baptist University, Hong Kong, China
- Siqin Wang** Spatial Sciences Institute, University of Southern California, Los Angeles, CA, United States
- Zhuosen Wang** Earth System Science Interdisciplinary Center, University of Maryland College Park, College Park; Terrestrial Information Systems Laboratory, NASA Goddard Space Flight Center Greenbelt, MD, United States
- Guanzhou Wei** Department of Industrial and Manufacturing Systems Engineering, Iowa State University, Ames, IA, United States
- Yixin Wen** Department of Geography, University of Florida, Gainesville, FL, United States
- Di Wu** Department of Geography, University of Colorado Boulder, Boulder, CO, United States
- Min Xu** Department of Geospatial Informatics, Troy University, Troy, AL, United States
- Yingwei Yan** Department of Geography, National University of Singapore, Singapore
- Zhengxiao Yan** Department of Geography and Environment; Carolina Population Center, University of North Carolina at Chapel Hill, Chapel Hill, NC, United States
- Di Yang** Department of Geography, University of Florida, Gainesville, FL, United States
- He Yin** Department of Geography, Kent State University, Kent, OH, United States
- Kunxiaojuan Yuan** Department of Earth & Atmospheric Sciences, University of Houston, Houston, TX, United States
- Chang Zhao** Agronomy Department, University of Florida, Gainesville, FL, United States
- Qunshan Zhao** School of Social and Political Sciences, University of Glasgow, Glasgow, United Kingdom
- Yilun Zhao** Department of Geography and Geographic Information Science, University of Illinois Urbana-Champaign, Urbana, IL, United States
- Zhijie Zhou** Department of Geography and Geographic Information Science, University of Illinois Urbana-Champaign, Urbana, IL, United States
- Siyu Zhu** School of Civil Engineering and Environmental Science, University of Oklahoma, Norman, OK, United States
- Alexander Zipf** GIScience Research Group, Institute of Geography, Heidelberg University; HeiGIT gGmbH at Heidelberg University, Heidelberg, Germany