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Supplemental information

**Social media data for environmental
sustainability: A critical review
of opportunities, threats, and ethical use**

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Table S1. Classification of the 415 studies in the database based on application and related Sustainable Development Goal target

SDG	SDG target	Application	Focus of supporting studies
	Sustainable agriculture (2.4)	<ul style="list-style-type: none"> Assess cultural significance of agricultural landscapes Map and monitor urban farming 	<ul style="list-style-type: none"> In Tuscany, Italy¹⁻³ and Europe⁴ Mountain agriculture in Trento, Italy⁵ Four world metropolises⁶
	Sustainable development education (4.7)	<ul style="list-style-type: none"> Explore meaning-making about environment and sustainability 	<ul style="list-style-type: none"> Young adults in Sweden⁷ Empathy expression in New York⁸
	Safe drinking water (6.1)	<ul style="list-style-type: none"> Gauge public attitude toward management Inform operation of supply systems 	<ul style="list-style-type: none"> Shutdown of water supply due to pollution in USA⁹ Water charges in Ireland¹⁰ Analyze water discourses¹¹ Catchment hydrology in Italian Alps¹²
	Water quality (6.3)	<ul style="list-style-type: none"> Promote sustainable sanitation Value benefits of quality improvement 	<ul style="list-style-type: none"> Benefits of nature-based solutions^{13,14} Recreation in Minnesota lakes¹⁵
	Water-related ecosystems (6.6)	<ul style="list-style-type: none"> Inform ecosystem restoration Assess provision of cultural ecosystem services Monitor environmental change 	<ul style="list-style-type: none"> Great Lakes, USA^{16,17} Tourism in a Ramsar wetland in India¹⁸ Wetlands in South Korea^{19,20}, India²¹, France²², Canada²³, China²⁴ Danube²⁵, Ebro^{26,27} river deltas Rivers in Idaho²⁸, the Netherlands²⁹ Lakes in the USA³⁰, globally³¹ Water level in a Saudi Arabian cave³²
	Access to energy services (7.1)	<ul style="list-style-type: none"> Explore public perception of energy 	<ul style="list-style-type: none"> Energy utilities merger in the Carolinas³³ Nuclear energy risk communication in the USA³⁴

	supply and energy projects	<ul style="list-style-type: none"> • Coal seam gas project in Alaska³⁵ • Debate concerning Keystone XL pipeline³⁶ • Hydraulic fracturing³⁷ • Communication by anti-coal activist movements³⁸ 	
	Share of renewable energy (7.2)	<ul style="list-style-type: none"> • Understand public opinion on renewable energy • Three Gorges Dam hydroelectricity project³⁹ • Preferences for energy policies in Spain and UK⁴⁰ • Local opposition to wind power project in Germany⁴¹ 	
	Sustainable tourism (8.9)	<ul style="list-style-type: none"> • Characterize spatial-temporal patterns of tourist visits • Analyze tourist movements and choices 	<ul style="list-style-type: none"> • Tourism in China⁴², Shenzhen⁴³, Hong Kong^{44,45}, Beijing⁴⁶⁻⁴⁸, Qingdao⁴⁹, Huangshan⁵⁰, Cilento, Italy⁵¹, Tokyo⁵², NE Portugal⁵³, India⁵⁴, Europe⁵⁵, worldwide^{56,57} • Hotspots in Nepal⁵⁸, European cities^{59,60} • Areas of Interest in six world metropolises⁶¹ • Tourism routes in New York City^{62,63} • Tourists' behavior in California^{64,65} • Visitors perceptions of lake tourism⁶⁶
	Sustainable and clean industries (9.4)	<ul style="list-style-type: none"> • Analyze sustainability marketing communication 	<ul style="list-style-type: none"> • Environmental disclosure of Brazilian companies⁶⁷ • Fortune 500 enterprises⁶⁸
	Inclusion (social, economic and political) (10.2)	<ul style="list-style-type: none"> • Address inequality in access to natural areas 	<ul style="list-style-type: none"> • Green gentrification in Barcelona⁶⁹ • Visitation and access in New York City's parks⁷⁰ • Inequality in access to protected areas in Chile⁷¹
	Public transport systems (11.2)	<ul style="list-style-type: none"> • Analyze cycling infrastructure and their use • Plan and improve public transport systems 	<ul style="list-style-type: none"> • Path networks in Belgium⁷², Amsterdam and Osnabrueck⁷³ • Investment in cycling infrastructure in Glasgow⁷⁴ • Exposure of cyclists to air pollution in Glasgow⁷⁵ • Cycling behavior in Glasgow⁷⁶ • Sentiment of transportation-related tweets during London Olympics⁷⁷ • Human mobility in Chicago⁷⁸ • Public opinion about transport system⁷⁹

Inclusive and sustainable urbanization (11.3)	<ul style="list-style-type: none"> • Characterize visual quality of urban landscape • Map urban functions and urban land use 	<ul style="list-style-type: none"> • City of Livorno, Italy⁸⁰, Province of Barcelona⁸¹, various cities⁸² • Public open spaces in Munich⁸³ • Amsterdam⁸⁴, London^{85,86}, Turku, Finland⁸⁷, Guangzhou, China⁸⁸⁻⁹⁰, Beijing⁹¹, Shanghai⁹², Edinburgh⁹³, Dakar⁹⁴, St. Petersburg⁹⁵, Sapporo⁹⁶, Chicago⁹⁷, Shenzhen⁹⁸, multiple cities^{99,100} • Monitor urbanization and urban sprawl¹⁰¹
Cultural and natural heritage (11.4)	<ul style="list-style-type: none"> • Examine use and management of heritage sites 	<ul style="list-style-type: none"> • Visitor flows worldwide¹⁰² • Tourist movement in Cuzco and Machu Picchu¹⁰³ • UNESCO World Heritage sites in conflict areas¹⁰⁴ • Historic urban landscape of Tripoli, Lebanon¹⁰⁵ • Communication about Marine World Heritage sites¹⁰⁶
Resilience to disasters (11.5)	<ul style="list-style-type: none"> • Detect and characterize flood extent and severity • Social sensing of natural hazards for footprint and damage assessment • Enhance preparedness, response and recovery 	<ul style="list-style-type: none"> • Flood extent maps in UK¹⁰⁷⁻¹⁰⁹, Germany^{110,111}, US^{112,113}, Philippines and Pakistan¹¹⁴, China¹¹⁵ • Prediction of flood events^{116,117} • Flood velocity and streamflow estimation^{118,119} • Flooding thresholds in the US East Coast¹²⁰ • Prioritization of flood response¹²¹ • Earthquakes^{122,123} • Hurricanes¹²⁴⁻¹²⁶ • Wildfires¹²⁷⁻¹²⁹ • High winds¹³⁰ • Heavy precipitation¹³¹ • Typhoon Haiyan¹³² • Hurricane Sandy¹³³⁻¹³⁷ • Wildfires^{138,139} • Crisis development, refugee flows after Arab Spring¹⁴⁰ • Multiple disasters, including Hurricane Sandy¹⁴¹ • Monitor community mood to enhance resilience¹⁴² • Severe weather risk communication¹⁴³

		<ul style="list-style-type: none"> • Winter storms¹⁴⁴ • Drought risk management¹⁴⁵ • Post-disaster (earthquake) recovery¹⁴⁶ 	
Urban green and public spaces (11.7)	<ul style="list-style-type: none"> • Assess ecosystem services of urban parks and green infrastructure • Evaluate well-being benefits from exposure to nature • Understand public opinion, perceptions and satisfaction 	<ul style="list-style-type: none"> • Green spaces in cities in Turkey^{147,148}, China¹⁴⁹⁻¹⁵⁷, Germany^{158,159}, USA¹⁶⁰, Canada¹⁶¹, Denmark¹⁶², Finland^{163,164}, Spain¹⁶⁵, Australia¹⁶⁶, Singapore¹⁶⁷⁻¹⁷⁰, UK^{171,172} • Places of importance for local urban residents¹⁷³ • Nanjing residents during COVID-19 pandemic¹⁷⁴ • Cities of Szeged, Hungary¹⁷⁵, San Francisco¹⁷⁶, Boston¹⁷⁷ • Urban parks in New York City¹⁷⁸, USA cities¹⁷⁹ • Green spaces in Dublin¹⁸⁰, London^{181,182}, Shenzhen¹⁸³, Beijing¹⁸⁴⁻¹⁸⁶, Birmingham^{187,188}, Seoul^{189,190}, New York City^{191,192}, Zurich¹⁹³ 	
 12 SUSTAINABLE CONSUMPTION AND PRODUCTION	Management of chemicals and wastes (12.4)	<ul style="list-style-type: none"> • Monitor solid waste management • Infer urban air pollution levels and inform air pollution debate 	<ul style="list-style-type: none"> • Odors from landfills in China¹⁹⁴ • Impact of litter on wildlife¹⁹⁵ • Power relations in communication about air pollution in China¹⁹⁶ • Predict particulate matter concentration in Beijing¹⁹⁷ • Atmospheric air quality and health effects, globally¹⁹⁸ • Air quality index for Chinese cities¹⁹⁹
	Corporate sustainable practices (12.6)	<ul style="list-style-type: none"> • Inform corporate sustainability practices • Assess impacts on business performance 	<ul style="list-style-type: none"> • Spillover effects of environmental regulation in China²⁰⁰ • Information from social media as input to decision-makers²⁰¹ • Assessment of supply chain risks and uncertainty²⁰² • Effect of social activism on stock market performance of Spanish banks²⁰³
	Sustainable development awareness (12.8)	<ul style="list-style-type: none"> • Uncover public perspectives on sustainability topics 	<ul style="list-style-type: none"> • Sentiment of sustainability-related tweets²⁰⁴ • Discourses on sustainability and consumption²⁰⁵ • Debate about seal hunting in Canada²⁰⁶ • Conceptions of health and sustainability in beverage consumption²⁰⁷ • Global debate about land grabbing²⁰⁸

	<ul style="list-style-type: none"> • Understand perceptions of nature • Explore spread of sustainability information • Analyze communication on environmental politics 	<ul style="list-style-type: none"> • Testing the biophilia hypothesis²⁰⁹ • Classifying public opinions on nature²¹⁰ • Discourse and dissemination around nature-deficit disorder²¹¹ • Effectiveness of science communication in Italy²¹² • Circulation of environmental information²¹³ • Biodiversity conservation awareness²¹⁴ • Communication about natural capital concept²¹⁵, conservation science²¹⁶, Deepwater Horizon oil spill²¹⁷ • Greens party in Australia²¹⁸ • UKIP party in UK²¹⁹
	<p>Resilience and adaptive capacity (13.1)</p> <p>Climate change policies (13.2)</p> <p>Climate change awareness (13.3)</p>	<ul style="list-style-type: none"> • Identify mismatches in socio-ecological systems • Assess impacts on nature-based recreation • Explore perception of impacts and policies • Analyze online discussions on climate change • Wildflower phenology and human visitation in Mount Rainier National Park²²⁰, green areas in Beijing²²¹ • Public response to heat waves²²² • National parks in the USA²²³ • Low-carbon cities in China²²⁴ • Remarkability of temperature anomalies²²⁵ • Discussions on COP21²²⁶, 2013 IPCC report²²⁷ • Gender differences²²⁸ and social network effects^{229,230} in climate change communication • Discourses of climate change skeptics or deniers²³¹⁻²³⁴ • Climate change-related discussions in China²³⁵, and globally²³⁶⁻²³⁸ • Effect of weather events on climate change discussion²³⁹⁻²⁴² • Climate change science learning and environmentally friendly behavior²⁴³
	<p>Marine and coastal ecosystems (14.2)</p>	<ul style="list-style-type: none"> • Assess coastal and marine ecosystem services • Whale watching in Sri Lanka²⁴⁴ • Recreation in NW Portugal²⁴⁵, Gold Coast, Australia²⁴⁶, Azov Sea shores²⁴⁷, beaches in Indonesia²⁴⁸ • Eco-tourism in Great Barrier Reef ^{249,250}, Seychelles²⁵¹ • Recreational fisheries^{252,253}

		<ul style="list-style-type: none"> • Global coral reef tourism²⁵⁴ • Multiple cultural ecosystem services in Lithuania²⁵⁵, Mexico²⁵⁶, Turkey²⁵⁷, the Netherlands²⁵⁸ 	
	<ul style="list-style-type: none"> • Map human interactions with marine species and disturbance to ecosystems 	<ul style="list-style-type: none"> • Cetacean occurrences in Italian Mediterranean Sea²⁵⁹ • Human interactions with Hawaiian monk seal²⁶⁰ • Environmental impacts from tourism in Iceland²⁶¹ • Environmental monitoring in Great Barrier Reef^{262,263} • Occurrence of jellyfish in Malta²⁶⁴ 	
Conservation of coastal areas (14.5)	<ul style="list-style-type: none"> • Assess benefits of marine protected areas 	<ul style="list-style-type: none"> • Multiple marine protected areas worldwide^{265,266} • Tarutao National Marine Park, Thailand^{267,268} • Coral Coast in Brazil²⁶⁹ 	
	Terrestrial and freshwater ecosystems (15.1)	<ul style="list-style-type: none"> • Map land use/land cover or geomorphometry • Analyze public perception and benefits of terrestrial protected areas • Investigate human-nature conflicts 	<ul style="list-style-type: none"> • Land use/land cover in urban green areas in London²⁷⁰, San Diego county, USA²⁷¹ • Landscape variation and folksonomies in Switzerland²⁷² • Geomorphometry in UK²⁷³ • Nature-based tourism and recreation in Spain^{274–276}, South Africa^{277–280}, Finland^{277,280–282}, Israel²⁸³, USA^{284–289}, Australia^{290,291}, UK²⁹², Germany^{293,294}, and globally^{295–297} • Landscape aesthetics in Yorkshire Dales National Park, UK²⁹⁸ • Public perceptions of national parks in Nepal²⁹⁹, South Africa³⁰⁰ • Mountain biking in Sintra-Cascais Natural Park, Portugal³⁰¹ • Multiple cultural ecosystem services of protected areas in Portugal^{302,303}, Spain^{304,305}, Argentina³⁰⁶, Finland³⁰⁷, Brazil³⁰⁸ • Tourism pressure in Korean national parks³⁰⁹ • Attractors for eco-tourists in sub-Saharan protected areas^{310–312} • Discourses about controversial environmental management issues in Australia³¹³ • Cattle grazing in Californian rangelands³¹⁴ • Hunting in British Columbia³¹⁵ • Unwanted visitors' behavior in South African national park³¹⁶

	<ul style="list-style-type: none"> Assess cultural ecosystem services 	<ul style="list-style-type: none"> Nature-based tourism and recreation in New Zealand³¹⁷, South Korea³¹⁸, UK³¹⁹, Florida³²⁰, Estonia³²¹, Switzerland³²², the Netherlands^{322,323}, Israel³²⁴, Costa Rica³²⁵, Argentina³²⁶, China³²⁷, Norway³²⁸, Europe^{329,330}, Arctic^{331,332}, Canada³³³, USA^{334–336} Wildlife watching in the USA³³⁷, Scotland³³⁸ Cultural ecosystem services in Germany³³⁹, multiple sites in Europe³⁴⁰, of native and non-native trees in Spain^{341,342} Experience tranquility^{343,344}
	<ul style="list-style-type: none"> Quantify landscape aesthetic values 	<ul style="list-style-type: none"> Visitors' perceptions in the UK^{345,346}, Spain³⁴⁷, USA^{348–350}, Switzerland^{351,352}, Turkey³⁵³, Estonia³⁵⁴, Slovenia³⁵⁵, Europe³⁵⁶, Japan³⁵⁷
	<ul style="list-style-type: none"> Complement traditional monitoring 	<ul style="list-style-type: none"> Species distribution, climate data, land use / land cover in Europe^{358–361} Population distribution^{362,363} Occurrence of landslides³⁶⁴ Vegetation phenology³⁶⁵ Index of environmental quality³⁶⁶ Seasonal color change in the environment³⁶⁷
	<ul style="list-style-type: none"> Evaluate environmental protection policies 	<ul style="list-style-type: none"> Benefits of investments in Public Land Acquisition³⁶⁸
Sustainable forest management (15.2)	<ul style="list-style-type: none"> Assess cultural ecosystem services of urban vegetation Assess cultural ecosystem services of forests 	<ul style="list-style-type: none"> Mangroves in Singapore^{369,370} Forests in Warsaw agglomeration³⁷¹ Forests in Tuscany, Italy³⁷² Mount Baker-Snoqualmie National Forest³⁷³ Gariwangsan and Yeoninsan forests in South Korea³⁷⁴ Forests on conserved lands in Vermont, USA³⁷⁵
Conservation of mountain ecosystems (15.4)	<ul style="list-style-type: none"> Assess benefits of mountain eco-tourism 	<ul style="list-style-type: none"> Indian Himalayan Region^{376,377} Berchtesgaden National park in Germany³⁷⁸ Dolomites UNESCO World Heritage Site³⁷⁹

		<ul style="list-style-type: none"> • Highest mountain in Australia^{380,381} • European Alps³⁸² • Mount Etna³⁸³
	<ul style="list-style-type: none"> • Assess mountain cultural ecosystem services 	<ul style="list-style-type: none"> • Dolomites UNESCO World Heritage Site³⁸⁴ • Aesthetic values in Austria³⁸⁵, France³⁸⁶ • 'Quatre montagnes' in the French Alps³⁸⁷
Loss of biodiversity (15.5)	<ul style="list-style-type: none"> • Collect information on species ecology and behavior 	<ul style="list-style-type: none"> • Spatial variation in species traits in Japan³⁸⁸ • Species determination in digital media³⁸⁹ • Ecology and behavior of shrikes³⁹⁰ • Winged ant emergence, autumnal house spider sightings, and starling murmurations across the UK³⁹¹ • Behavior of red and grey squirrels in Europe³⁹² • Trophic interactions³⁹³
	<ul style="list-style-type: none"> • Map species distribution 	<ul style="list-style-type: none"> • Flowering plants in the UK³⁹⁴ • Monarch butterfly and snowy owl³⁹⁵ • Bees and flowering plants in Australia³⁹⁶ • Iberian Argiope spider species in Spain³⁹⁷
	<ul style="list-style-type: none"> • Characterize human-wildlife interactions 	<ul style="list-style-type: none"> • African painted dog³⁹⁸ • Giant panda in China³⁹⁹ • Iconic terrestrial vertebrates in French alpine national parks⁴⁰⁰ • Birds in Chicago, IL, USA⁴⁰¹ • Tourism pressure in grizzly bear recovery areas⁴⁰² • Grouse species during winter recreational activities⁴⁰³
	<ul style="list-style-type: none"> • Monitor and respond to illegal activities 	<ul style="list-style-type: none"> • Illegal sport hunting in Brazil⁴⁰⁴ • Conservation-related violence toward poachers⁴⁰⁵
	<ul style="list-style-type: none"> • Analyze perceptions of biodiversity and endangered species 	<ul style="list-style-type: none"> • Sentiment toward iconic species⁴⁰⁶ • Global Important Bird and Biodiversity Areas⁴⁰⁷ • Public perception of slow lorises⁴⁰⁸ • Public engagement with endangered species⁴⁰⁹

Protected species trafficking (15.7)	<ul style="list-style-type: none">• Monitor online wildlife trade	<ul style="list-style-type: none">• Indonesian songbirds⁴¹⁰• Orchids⁴¹¹• Slow loris trade in Turkey⁴¹²
Invasive alien species (15.8)	<ul style="list-style-type: none">• Monitor spread of non-native species	<ul style="list-style-type: none">• Freshwater turtles in the UK⁴¹³• Oak processionary in Europe⁴¹⁴• Oak processionary moth, emerald ash borer, eastern grey squirrel⁴¹⁵

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