

Science of cloud and climate science: An analysis of the literature over the past 50 years

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Abstract

Clouds pose a particularly difficult challenge within Earth’s climate system. They are relatively small in spatiotemporal scale but still have a strong influence on radiative fluxes, global circulation, and precipitation patterns. Increasing research attention has been devoted to them over the past 50 years, and we give a summary of the resulting body of scientific literature in this introductory chapter. Articles on clouds and climate are doubling every 8 years, a rate about twice that of scientific publications generally. This expanding number of publications correlates with more citations, but citation rates have also slowed in the most recent decade, despite a growing number of atmospheric science students. We show some basic “science of science” (SciSci) analyses of the clouds and climate literature, such as authorship networks or abstract text mining for techniques, and suggest that further SciSci analyses may help us to process the proliferation of results on clouds and climate and optimize how we do research in the crucial years ahead.

1 **Science of cloud and climate science: An analysis of the**
2 **literature over the past 50 years**

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6 **Summary**

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 8 relatively small in spatiotemporal scale but still have a strong influence on radiative fluxes,
 9 global circulation, and precipitation patterns. Increasing research attention has been de-
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 16 climate literature, such as authorship networks or abstract text mining for techniques,
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 18 sults on clouds and climate and optimize how we do research in the crucial years ahead.

19 **1 Research on Clouds and Climate**

20 Clouds have a multi-faceted impact on the climate. They affect the terrestrial radi-
 21 ative balance by reflecting visible and ultraviolet radiation from the sun and absorb-
 22 ing infrared radiation from the Earth surface. They are also coupled to the large-scale
 23 circulation. Cloud formation is determined by where circulation patterns bring moisture
 24 aloft, while the reflection and absorption of radiation by clouds also feeds back on cir-
 25 culation. Finally, precipitation is generated from clouds. The intensity, frequency, and
 26 duration of precipitation are determined by cloud structure and dynamics.

27 Climatic impacts of clouds are also multi-scale, involving a huge range of processes,
 28 from new particle formation at the nanometer scale up to atmospheric wave propaga-
 29 tion over thousands of kilometers. Simulating or measuring these 15 orders of magni-
 30 tude challenges our computational and observational tools. And with phenomena as di-
 31 verse as turbulence and nucleation, and forms as varied as cumulonimbus towers and stra-
 32 tocumulus decks, clouds challenge our ability to simplify or generalize.

33 This combination of impact and complexity has piqued the interest of ever more
 34 researchers and funding agencies and given rise to an imposing body of scientific liter-
 35 ature. As in other scientific fields, this rapid growth in publication has motivated sys-
 36 tematic review and meta-analysis, in this case through entities like the Intergovernmen-
 37 tal Panel on Climate Change (IPCC) or the Climate Model Intercomparison Projects
 38 (CMIP). The fifth assessment report of the IPCC contained a chapter dedicated to clouds
 39 and climate (Boucher et al., 2013), and the most recent report devotes sections to cloud
 40 feedbacks and water cycle changes with warming (Forster et al., 2021; Douville et al.,
 41 2021). Within the most recent CMIP experimental design is an endorsed model inter-
 42 comparison project focused exclusively on cloud feedbacks (CFMIP) (Webb et al., 2017).

43 This monograph represents one of such increasingly important community-wide re-
 44 view efforts. With primarily early-career lead authors, each chapter provides a review
 45 of the existing scientific literature and open questions in a given subfield. The volume
 46 is intended to act as a resource for graduate students, both to orient in their new sub-
 47 field and gain fluency in related ones. For young scientists, it may direct their energy
 48 toward high-impact questions or help them formalize their future research program, and
 49 for more established scientists, it may generate ideas for collaboration.

50 Chapters have been organized around the climatic impacts of clouds on radiation,
 51 circulation, and precipitation described above (Fig. 1). Chapters 2 through 4 focus on
 52 the formation of liquid droplets and ice crystals on particulates and how this cloud for-
 53 mation can affect radiative fluxes, both at the surface and in the atmosphere. Chapters
 54 5 through 9 review research on the most climatically-relevant cloud types including Arc-
 55 tic mixed-phase, extratropical, tropical marine low, and tropical organized deep. Cloud-

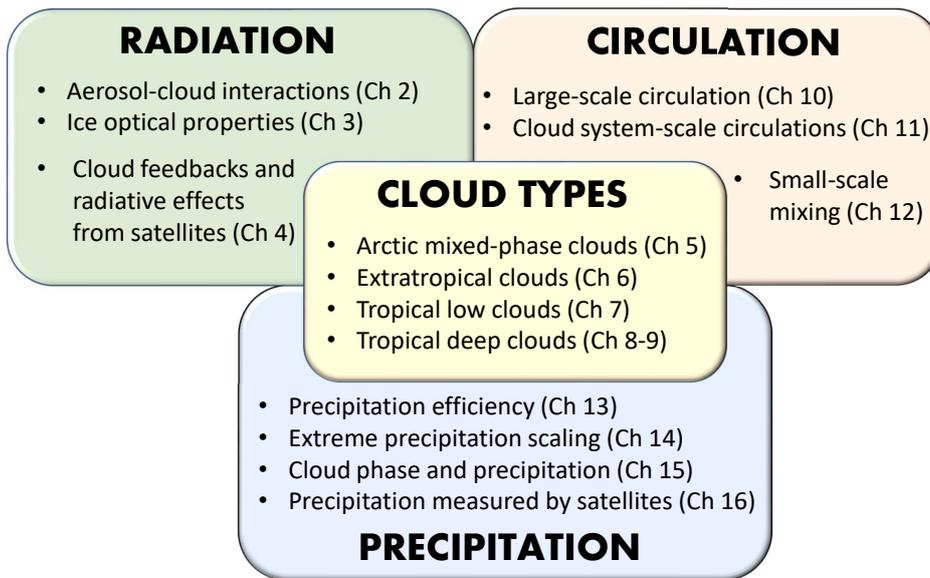


Figure 1. The monograph chapters are organized around climatic impacts of clouds on radiation, circulation, and precipitation.

56 circulation coupling at the global, meso, and micro scales are covered in Chapters 10 through
 57 12. The final monograph section discusses precipitation efficiency, extremes, phase, and
 58 measurements (Chapters 13-16).

59 **1.1 Science of Science for Clouds and Climate**

60 As an imposing body of scientific literature might already suggest, clouds and cli-
 61 mate research at the institute or university level has tended to answer questions by per-
 62 forming new simulations or analyses, rather than by synthesizing output from existing
 63 studies. There could be many possible reasons for this focus on generating new output,
 64 for example limited code and data documentation and accessibility in the past or cost
 65 and difficulty of storing and analyzing large volumes of data. In any case, systematic re-
 66 view at the community level is fueled by this very large number of individual studies.
 67 In this introductory chapter, we posit that analyzing how these individual studies are
 68 produced in a kind of “science of science” is another meta-exercise that would be use-
 69 ful for the clouds and climate community.

70 “Science of science” (SciSci) is an emerging area of study that combines sciento-
 71 metrics and the sociology of science to understand and optimize how science is done, from
 72 the emergence of new paradigms to the career path of students. As Fortunato et al. (2018)
 73 note in their review, SciSci has been driven in recent years by increasingly quantitative
 74 and accessible data on publications in databases like Scopus or Web of Science and by
 75 collaboration of natural, computational, and social scientists. Although some SciSci find-
 76 ings are domain- or culture-specific, a number of generalizable statements can be made.
 77 For example, networks of scientific concepts, tools, and authors tend to densify over time,
 78 indicating risk aversion and the tendency to select questions and collaborations conser-
 79 vatively (Fortunato et al., 2018; Foster et al., 2015). Such densification can be danger-
 80 ous, as a small subset of authors cite one another and reinforce established hypotheses
 81 in an echo chamber effect.

82 SciSci analyses have not yet been done for clouds and climate research to our knowl-
 83 edge. Two decades ago, Geerts (1999) aggregated data on atmospheric science publica-
 84 tions and found an increasing number of journals, pages per journal, and words per page
 85 in articles published between 1965 and 1995. But we have not found studies building upon
 86 this one or studies focused exclusively on climatic impacts of clouds. As a field with bur-
 87 geoning student interest and the pressure to inform climate policy reliably and rapidly,
 88 clouds and climate research could benefit from such meta-study. To catalyze these ef-
 89 forts, we perform preliminary SciSci analyses in this introductory chapter.

90 1.2 Publication Data and Methods

91 We draw our publication data from the Scopus database of Elsevier, which con-
 92 tains abstracts and citations of peer-reviewed academic literature. A set of boolean key-
 93 words define the Scopus query for publications on clouds and climate, as well those fil-
 94 tered by theme (Tab. 1). For example for all publications in the field, we require that
 95 the title, abstract, or keywords contain both the term “clouds” and the term “climate”
 96 (TITLE-ABS-KEY(“clouds” AND “climate”). We also require that the document be
 97 an English language journal article past the review stage (LIMIT TO(PUBSTAGE(“final”))
 98 AND LIMIT-TO(DOCTYPE, “ar”) AND LIMIT-TO(LANGUAGE, “English”)) and that
 99 the journal domain be Earth or environmental sciences (LIMIT-TO(SUBJAREA(“EART”)
 100 OR LIMIT-TO(SUBJAREA(“ENVT”))). We find a non-negligible number of publications
 101 on Mars for some queries and additionally omit these (NOT TITLE-ABS-KEY(“Mars”)).

102 Scopus queries are also used to classify techniques by searching all clouds and cli-
 103 mate abstracts for keywords. For example, we classify abstracts that contain the strings
 104 “model”, “parameterization”, “simulation”, “GCM”, “trajector”, or “radiative transfer”
 105 as modelling work. Abstracts can be classified as employing multiple techniques if they
 106 satisfy multiple queries. The anonymous Author ID field of Scopus queries is used with
 107 the Python networkx package to generate authorship networks. Finally, the titles asso-
 108 ciated with some queries are used to generated word clouds, using the Python WordCloud
 109 package. WordCloud takes the title text and fills space with individual words, sized by
 110 their frequency. A set of default “stop words” is omitted (listed in https://github.com/amueller/word_cloud/blob/master/wordcloud/stopwords) to which we add the fol-
 111 lowing words: *cloud*, *climate*, *using*, *change*, *comparison*, *evaluation*, *effect*, *global*, *study*,
 112 *effects*, *atmospheric*, *changes*, *response*, *based*, *part*, *characteristic*, and *influence*. By elim-
 113 inating these words, concrete article themes and techniques emerge more clearly from
 114 the title text.
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116 The queries in Tab. 1 are denoted **Clouds and Climate** for all publications on
 117 clouds and climate; **Impacts** for publications decomposed into cloud impacts on radi-
 118 ation, circulation, and precipitation corresponding to the monograph sections; **Chap-**
 119 **ters** for publications decomposed according to the monograph chapters; and **Techniques**
 120 for identification of the techniques used in the abstracts of the clouds and climate pub-
 121 lications. With the queries in Tab. 1, advanced searches can be reproduced at www.scopus.com/search/form.uri?display=advanced. Publication data for the figures has been
 122 downloaded on 3 September 2020 and is shown only through the end of 2019. The only
 123 non-Scopus data used concerns doctorates granted and funding awarded for the United
 124 States, which we take from the National Center for Science and Engineering Statistics
 125 (NCSES) Survey Data at <https://ncesdata.nsf.gov/home>. For US doctorate degrees
 126 granted in geosciences, we look at the Survey of Earned Doctorates at <https://ncesdata.nsf.gov/builder/sed> with Doctorate Recipients as our measure and Academic Dis-
 127 cipline: Geosciences, atmospheric sciences, and ocean sciences as our dimension. For US
 128 annual funding, we look at the Survey of Federal Funds for Research and Development
 129 at <https://ncesdata.nsf.gov/builder/ffs> with Research Obligations as our mea-
 130 sure and Fields of Study: Atmospheric Science as our dimension.
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Table 1. Boolean syntax for Scopus database queries. We filter for English-language journal articles (“ar”) in the finalized publication stage (“final”), classified as Earth and Planetary Sciences (“EART”) or Environmental Sciences (“ENVI”). Filters other than TITLE-ABS-KEY are held constant for all searches and denoted *other filters*.

<i>Queries</i>	
Clouds and Climate	TITLE-ABS-KEY(“clouds” AND “climate”) AND NOT TITLE-ABS-KEY(“Mars”) AND LIMIT-TO(PUBSTAGE(“final”)) AND LIMIT-TO(DOCTYPE,“ar”) AND LIMIT-TO(SUBJAREA,“EART”) OR LIMIT-TO(SUBJAREA,“ENVI”) AND LIMIT-TO(LANGUAGE,“English”)
Impacts + <i>other filters</i>	TITLE-ABS-KEY(“clouds” AND “climate” AND “circulation”) TITLE-ABS-KEY(“clouds” AND “climate” AND “radiati*”) TITLE-ABS-KEY(“clouds” AND “climate” AND “precipitation”)
Chapters + <i>other filters</i>	TITLE-ABS-KEY(“equilibrium climate sensitivity” OR “cloud feedback*”) TITLE-ABS-KEY((“radiative transfer” AND “climate” AND “cloud”) OR “cloud radiative effect*”) TITLE-ABS-KEY(“cloud classification”) TITLE-ABS-KEY((“cloud microphysics” OR (“cloud” AND “microphysics”) OR “microphysics parameterization”) TITLE-ABS-KEY((“aerosol indirect effect” OR “aerosol-cloud interaction”) TITLE-ABS-KEY((“atmospher*” AND (“dynamical core” OR “primitive equations”)) TITLE-ABS-KEY(“radiative-convective equilibrium” OR “convective organization” OR “convective aggregation” OR “organized convection”) TITLE-ABS-KEY(“cloud-circulation coupling” OR (“cloud” AND “large-scale circulation”)) TITLE-ABS-KEY(“cloud” AND “field campaign”) TITLE-ABS-KEY(“cloud” AND “ground-based measurement”) TITLE-ABS-KEY((“cloud” AND “machine learning”) OR (“cloud” AND “causal inference”))
Techniques	ABS(“in-situ” OR “flight” OR “campaign” OR “aircraft” OR “rocket” OR “drone”) ABS(“model” OR “parameterization” OR “simulation” OR “GCM” OR “trajector” OR “radiative transfer”) ABS(“reanalys” OR “emission”) ABS(“satellite” OR “CERES” OR “TRMM” OR “ISCCP” OR “remote sens” OR “retrieval” OR “imager” OR “CALIPSO” OR “CloudSat” OR “MODIS” OR “mission”) ABS(“ground-based” OR “station” OR “meteorological observator” OR “rain gauge” OR “site” OR “SHEBA” OR “flux tower”) ABS(“laboratory” OR “chamber” OR “chemical characteriz”) ABS(“observation”)

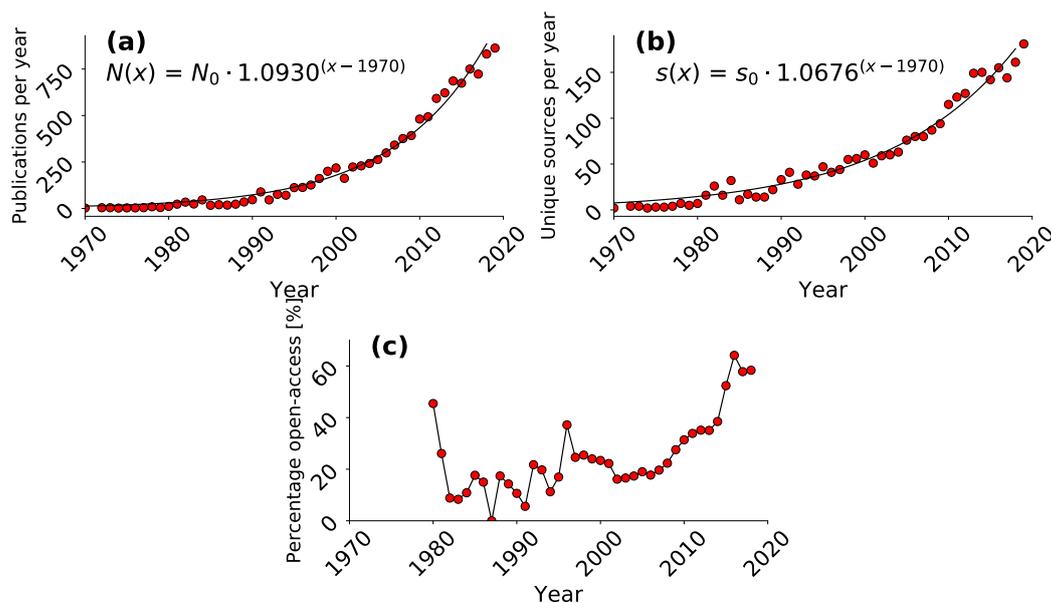


Figure 2. Atmospheric science publications, particularly open-access ones, and their unique sources have all increased dramatically over the past 50 years. Total number of publications per year from 1970 through 2019 with the **Clouds and Climate** query in Tab. 1 and their exponential fit (panel a). Number of unique journals for these publications per year from 1970 through 2019 and their exponential fit (panel b). Percentage of publications that have been open-access over time (panel c).

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2 Publications on Clouds and Climate

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We quantify the rapid growth of interest in clouds and climate with the output of the Scopus **Clouds and Climate** query. We find that the number of articles published on clouds and climate has been doubling every 8 years since 1970 (Fig. 2a). While an average of 27 articles per year were published in the 1980s, this rate had increased more than ten-fold by the 2000s. In the 2010s, an average of 660 articles per year were published. By comparison, Price (1963) and Fortunato et al. (2018) find a 15-year doubling period for scientific articles more generally, while Milojević (2015) cites 16- and 19-year doubling periods for physics and biomedical articles respectively.

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We could also consider how this growth compares to that of other subfields in climate science. Searching instead for publications on *ocean and climate* or *air pollution and climate*, these have doubling times of 7.9 and 8.5 years respectively, comparable to that of clouds and climate (not shown). Publications on the *biosphere and climate* are growing somewhat more gradually with a doubling time of 8.9 years, while publications on the *land surface*, the *cryosphere*, or the *carbon cycle* have stronger recent growth relative to cloud research, doubling in only 6.6 years, 5.1 years, and 6.0 years respectively. These doubling rates show that publication growth for climate research is about twice as fast as that for general scientific research.

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Not only is publication number on clouds and climate rapidly rising, the number of unique journals has also increased over time, doubling roughly every 11 years (Fig. 2b). In the 1980s, with the focus more often on meteorology, the *Journal of Atmospheric Sciences* published the most articles. By the 1990s, spurred by the release of the Charney Report and the launch of early geostationary satellites, interest broadened to the

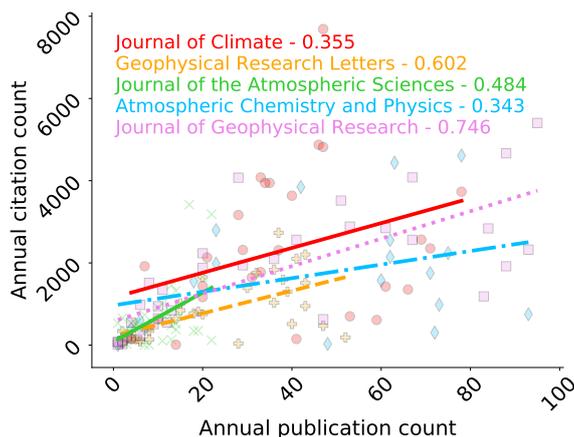


Figure 3. Publication and citation number are positively correlated, but not always strongly. Annual publication counts from 1970 through 2019 scattered against annual citation counts over the same period for the five most prolific atmospheric science publications; *Journal of Climate* and *Journal of the Atmospheric Sciences* are shown as solid traces, *Geophysical Research Letters* as a dashed trace, *Atmospheric Chemistry and Physics* as a dotted-dashed trace, and *Journal of Geophysical Research* as a dotted trace. Correlation coefficients are given next to the publication name in the legend.

156 role of clouds in climate with more publications now in the *Journal of Climate* and *Jour-*
 157 *nal of Geophysical Research*. Although only initiated in 2001, *Atmospheric Chemistry*
 158 *and Physics (ACP)* already published the most articles by the 2010s. As an open-access
 159 publisher, *ACP* output has also driven the large increase in open-access publication per-
 160 centage over the past 4 decades from less than 10% in the 1980s to more than 50% in
 161 the late 2010s (Fig. 2c).

162 2.1 Citation and Readability

163 Does this expanding body of literature correspond to higher readership and, hence,
 164 citation? For the five most prolific journals publishing on clouds and climate, higher an-
 165 nual publication counts do correlate with higher annual citation counts from 1970 through
 166 2019, but not always very strongly (Fig. 3). In spite of – or perhaps because of – its open-
 167 access policy, high publication rates in *ACP* correlate least strongly with high citation
 168 rates. From linear fits to these publication-citation scatter plots, annual citation count
 169 per journal increases by 33 for each additional article published. This annual citation
 170 increase per publication is largest for the *Journal of Atmospheric Sciences* at almost 60.

171 Without filtering for journal, cumulative distributions show that cloud and climate
 172 citation peaks for articles written between 1995 and 2005 (Fig. 4a-b). We may expect
 173 decreasing citation rates for more recent years, as the time since publication shortens,
 174 but the decreasing citation rates prior to 1995 indicate that older literature has been cited
 175 less than more recent literature. In the late 1990s, almost 20% of articles have more than
 176 a hundred citations, and only 10% are cited five times or fewer, counter to the highly-
 177 skewed nature of most citation distributions in other fields in which many articles are
 178 never cited and a very small number are very highly cited (Price, 1965).

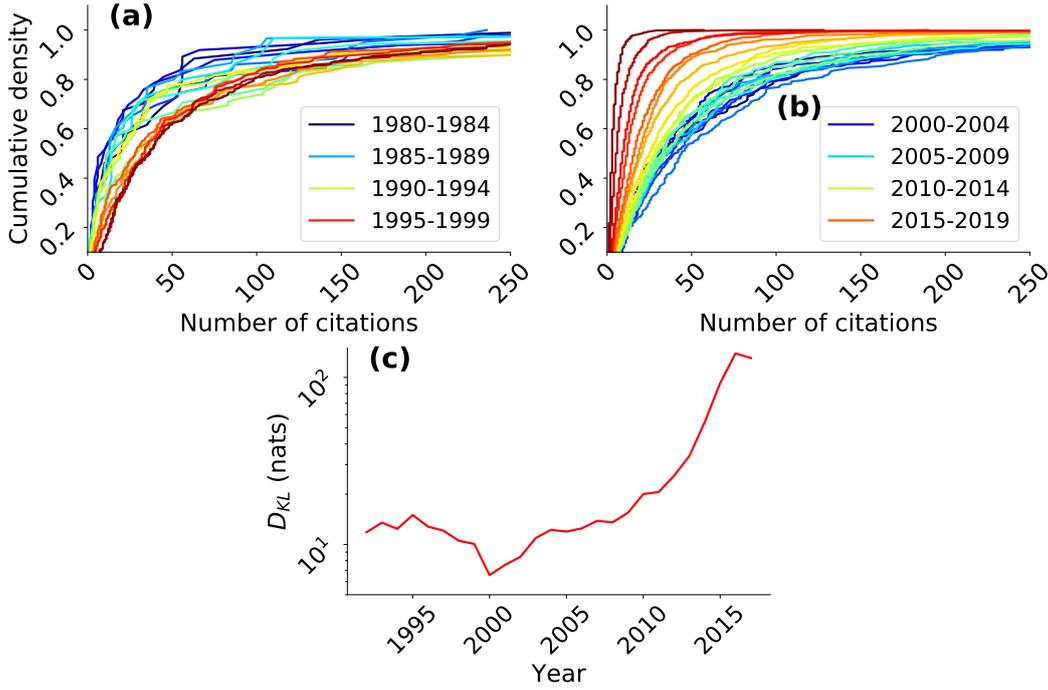


Figure 4. Citation rates in clouds and climate research have slowed relative to publication rate in the most recent decade. Total number of citations per article per year from 1980-1999 (panel a) and from 1999-2019 (panel b). Time series of the Kullback-Leibler divergence between the cumulative density of citations in a given year and the preceding year (panel c).

To further quantify citation trends, we employ the Kullback-Leibler divergence:

$$D_{KL}(P||Q) = \sum_{x_i} P(x_i) \log_2 \left(\frac{P(x_i)}{Q(x_i)} \right) \quad (1)$$

179 where P and Q are two probability distribution functions of annual citation numbers,
 180 x_i . The larger the value of D_{KL} , the more different the distributions P and Q are. Stated
 181 more formally, the larger the value of D_{KL} , the more information (in nats when D_{KL}
 182 is evaluated with the natural logarithm) would be lost in replacing P by Q . D_{KL} is not
 183 symmetric, and we always take the preceding year as the reference distribution (Q). Cal-
 184 culating a 5-year running mean of this pairwise D_{KL} yields relatively stable values of
 185 10 to 15 until 2007; thereafter, D_{KL} increases monotonically up to a most recent value
 186 of 140 (Fig. 4c). These increasing D_{KL} values indicate that citation growth in recent
 187 years is not maintaining pace with publication growth.

188 Decreased readability could contribute to this seeming drop in readership in recent
 189 years. The assessment of atmospheric science literature in Geerts (1999) found an in-
 190 creasing number of journals, pages per journal, and words per page in articles published
 191 between 1965 and 1995. While we reproduce this increasing trend in article length, page
 192 number per article, as a crude metric of readability, has been growing far less rapidly than
 193 publication or journal numbers, at an average of only a page per decade since 1970 (Fig.
 194 5a). Geerts (1999) also proposed that lagging US federal funding and the plateauing num-
 195 ber of PhD students could slow these publication rates in coming years. We find instead
 196 that, while US funding has converged or even dropped in recent years, the number of US
 197 students in geosciences has continued to grow (Fig. 5b). Equivalent data are not as read-

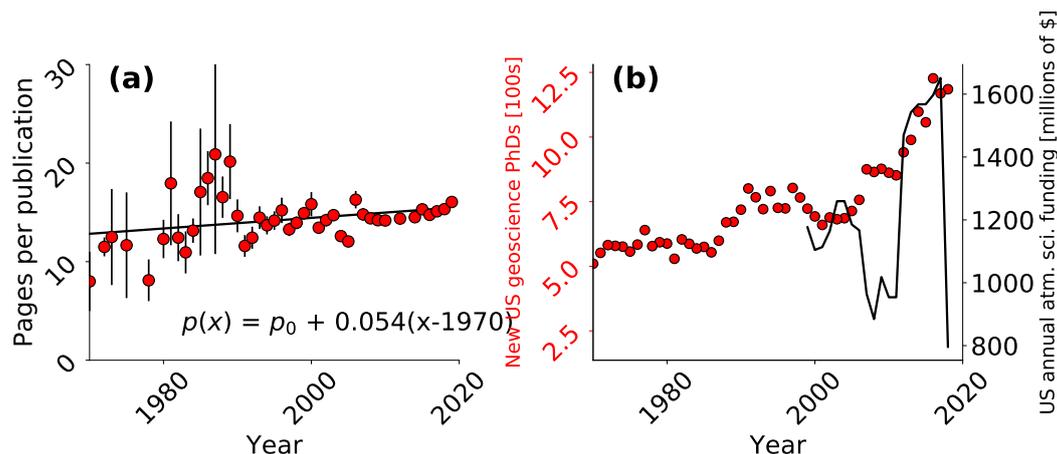


Figure 5. Articles have become only slightly less readable with page number as a metric, while potential readership has doubled since 2000 with graduating doctorates as a metric. Average number of pages per publication per year including standard error and a linear fit in black (panel a). US doctoral degrees awarded in geosciences, including both atmospheric and oceanic sciences, and US federal funding awarded to atmospheric sciences projects (panel b) with data for the last panel come from the US National Center for Science and Engineering Statistics Survey of Earned Doctorates and Survey of Federal Funds for Research and Development (see Sec. 1.2).

198 ily available for other regions, however, and these US trends may not be representative
 199 globally.

200 **3 The Role of Clouds in Radiation, Circulation, and Precipitation**

201 To understand how research effort has been devoted across the climatic impacts
 202 of clouds, we next decompose the publication trends from Section 2, using the **Impacts**
 203 queries from Table 1. Publications concerning radiation have been most numerous in the
 204 past three decades (Fig. 6a), averaging approximately 50 per year in the 1990s, more than
 205 twice that many in the 2000s, and more than five times that many in the 2010s. In word
 206 clouds of the titles from these publications on clouds and radiation (Sec. 1.2), studies
 207 of aerosol and using simulations are dominant, and the most commonly studied regions
 208 are the Arctic and the tropics (Fig. 6b). Feedback quantification and the surface energy
 209 budget also emerge as common topics.

210 Publications concerning precipitation have increased most rapidly among the three
 211 impacts, doubling every 7 years since 1970 (Fig. 6a). In the associated title word clouds,
 212 model simulation and satellite observation appear about equally, and the Arctic and the
 213 tropics emerge again as important areas of study, as does China (Fig. 6c). Influence of
 214 aerosol and convection or convective parameterization also appear prominently. Perhaps
 215 because of the computational demand associated with scale separation, growth has been
 216 slowest for publications on clouds and circulation with a doubling time of 10 years. Even
 217 more than cloud-radiation studies, these are heavily simulation-oriented (Fig. 6d). Aerosol,
 218 feedbacks, and surface interactions are again frequent topics, but less so than for clouds
 219 and radiation.

220 We can further decompose these trends into topics covered in monograph chapters
 221 (Fig. 7, **Chapters** queries in Tab. 1). Cloud microphysics is the most active subtheme

222 with almost 400 publications in 2019 alone (Fig. 7 inset). Its growth was particularly
 223 rapid in the early to mid-1990s, increasing seven-fold over six years. Precipitation mea-
 224 surement has also been a very active area of research since the mid-1990s with 160 pub-
 225 lications in 2019 alone. Slow and steady progress over the past three decades has char-
 226 acterized the optimization of dynamical cores and extension of ground-based measure-
 227 ments. Research output on cloud-circulation coupling has also been more gradual, again
 228 possibly because of the computational challenge to simulate all involved scales. Most dra-
 229 matic is recent growth in machine learning work. While almost no studies used machine
 230 learning in 2012, there were 200 publications in 2019 alone.

231 4 Methodology in Clouds and Climate

232 4.1 Techniques

233 The word clouds in Fig. 6b-d reveal some of the most common techniques used in
 234 clouds and climate research. We can organize these techniques around pillars of mod-
 235 elling, in-situ or ground-based measurement, and satellite retrieval, and we next search
 236 the clouds and climate abstract text in the **Techniques** queries of Tab. 1 (Sec. 1.2).

237 With this classifications of abstracts, almost 70% mention some kind of modelling
 238 (Fig. 8a). Coarse-resolution global climate models (GCMs) still dominate these mod-
 239 elling abstracts (45.8%), although high-resolution or cloud-resolving models account for
 240 another 10% each (Fig. 8b). Radiative transfer calculations make up another 18%, while
 241 less than 10% of abstracts are attributed to each of large-eddy simulation, trajectory anal-
 242 yses, and idealized or parcel simulations.

243 But modelling alone characterizes only 22% of abstracts; it is more often used in
 244 combination with satellite, flight, reanalysis, or ground-based data (Fig. 8c). Similarly,
 245 almost 40% of studies use remote sensing products, but only 6% use remote sensing prod-
 246 ucts exclusively. Ground-based measurements appear in 25% of abstracts, and flight mea-
 247 surements, given their more limited duration, appear in 8%. Laboratory studies are the
 248 most infrequent at only 3%, although these abstracts are also hardest to classify with
 249 their varied methods and keywords. While models and satellite data provide good spa-
 250 tiotemporal coverage and robust statistics for studies, we conclude that clouds and cli-
 251 mate research is fundamentally driven by a synergy of different techniques.

252 4.2 Authorship

253 Along with publication trends and techniques, we can also examine the personal
 254 aspect of how clouds and climate research is done. Do certain authors always work with
 255 the same coauthors? Do two or three “author clusters” characterize certain subfields?
 256 We use anonymous Author IDs from the **Impacts** and **Chapters** searches to generate
 257 author networks. Circular networks visualize the increase in authorship complexity over
 258 time dramatically; in such networks, nodes represent authors, edges represent co-authorship,
 259 and node color indicates degree, or coauthor number in our case.

260 We give an illustrative example of coauthorship evolution in the circular network
 261 of Fig. 9, taken from the **Impacts** query on clouds and radiation. In 1989, this network
 262 contained only 21 nodes, or authors, with an average degree, or coauthor number, of 1.52.
 263 A decade later in 1999, the same network contained 294 authors with an average coau-
 264 thor number of 3.74. Author number increases by more than an order of magnitude for
 265 the other **Impacts** queries in the decade between 1989 and 1999 also (not shown). Au-
 266 thorship in the most recent decade is too complex to present in this layout, but by 2019,
 267 author numbers involved in clouds and climate research had increased more than 70-fold
 268 since 1989.

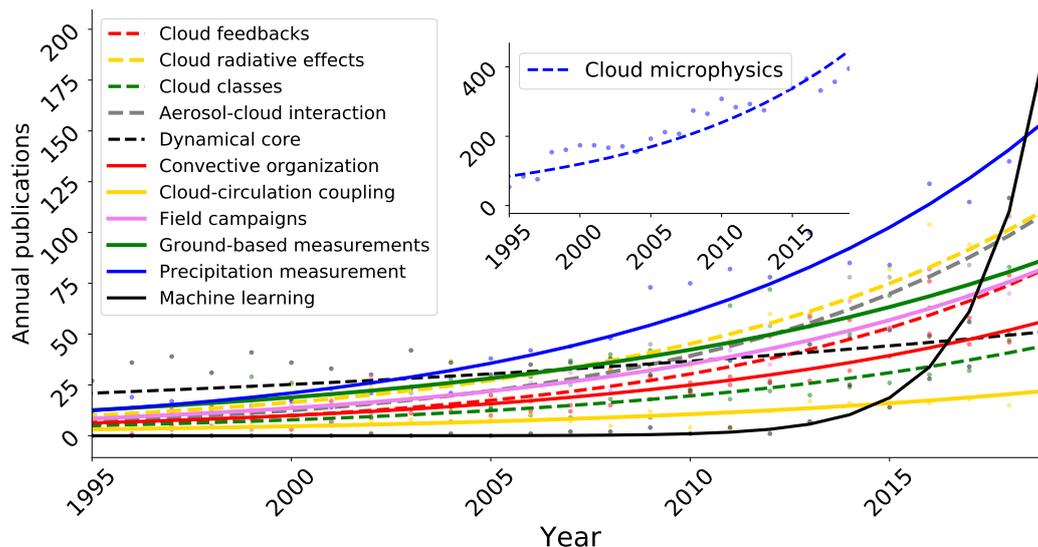


Figure 7. Recent growth in machine learning publication has been dramatic. Cloud microphysics and precipitation measurement are also very active areas of research. Publication trends decomposed into chapter topics, as well as their exponential fits, based on the Scopus **Chapters** queries. The inset y-axis for microphysics publications is two times larger than the primary y-axis.

269 We can describe networks not only with their degree, but also with their density.
 270 Network density refers to the degree normalized by the maximum possible number of edges,
 271 or the absolute coauthor number relative to its possible maximum value. Looking again
 272 at Fig. 9, while absolute co-authorship increased, relative co-authorship, as quantified
 273 by density, decreased six-fold between 1989 and 1999 for work on clouds and radiation.
 274 For work on clouds and precipitation, the drop in relative co-authorship is even more dra-
 275 matic at 30-fold between 1989 and 1999. In the two decades from 1999 to 2019, relative
 276 co-authorship dropped another two- to three-fold. The takeaway is that while those in
 277 clouds and climate research are collaborating more in absolute terms, they are also di-
 278 rectly involved in less and less of the total body of research produced by the community.

279 Increases in absolute co-authorship may suggest increased objectivity in our results,
 280 but as noted above, densification of author networks can actually have an opposite echo
 281 chamber effect in which author clusters reinforce their own hypotheses. This kind of clus-
 282 tering cannot be seen in a circular network, so we present authorship networks for the
 283 **Chapters** queries also in a spring layout (Fig. 10). A spring layout is generated by a
 284 force-directed algorithm in which edges are interpreted as springs governed by Hooke's
 285 Law, and nodes are arranged to minimize energy in the network.

286 These force-directed graphs for authorship tend to have a few studies with big groups
 287 of authors and many with smaller groups (Fig. 10). Ideally there would be some over-
 288 lap and not too much distance between the “hub” studies with large coauthorship and
 289 the “rim” studies with limited coauthorship. Work on cloud-radiative effects exhibits such
 290 a structure in which collaborations of different size are fairly well mixed, as does work
 291 on aerosol-cloud interaction and precipitation measurement, although to a lesser degree.
 292 In contrast, work on dynamical cores is characterized by a few studies with large author-
 293 ship and many smaller-scale efforts that do not overlap at all; however, such separation
 294 may be due to model-specific development in this case. We also see the cooperative na-

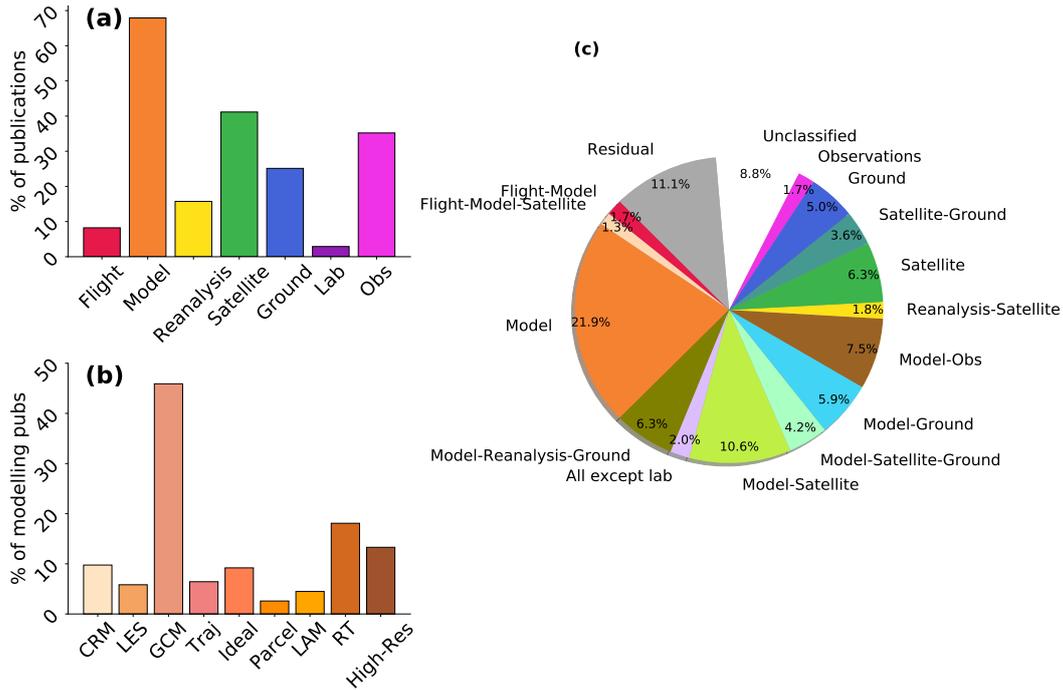


Figure 8. About 40% of studies use a combination of modelling and/or satellite data, but a great diversity of techniques are used to study clouds and climate. Percentage of techniques mentioned in **Clouds and Climate** abstracts, identified in the **Techniques** query (panel a). **Obs** designates unspecified observations, and studies on emissions inventories are grouped into the **Reanalysis** category. Additional filtering of modelling publications into model type (panel b). The following acronyms are used to label: Cloud-resolving model (CRM), global climate model (GCM), trajectories (Traj), idealized or RCE setups (Ideal), limited-area or regional modeling (LAM), radiative transfer calculations (RT), and high-resolution simulation (High-Res). Overall distribution of techniques or technique combinations (panel c). The generic Observations (Obs) classification is removed if another observational classification, e.g. satellite or ground-based was made.

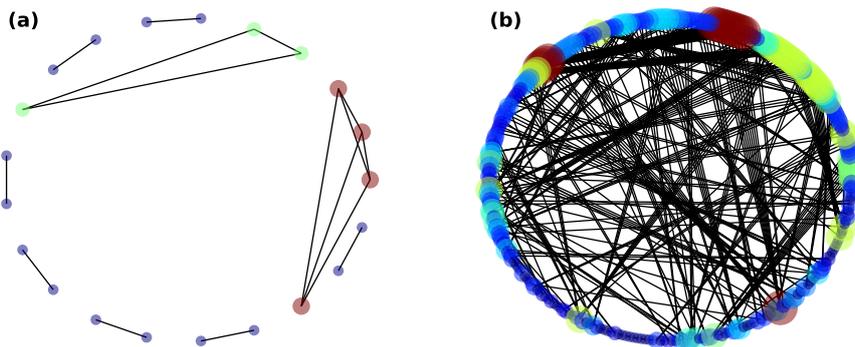


Figure 9. Over a decade, author number increases by a factor of 14 and the degree of co-authorship more than doubles. Evolution of author networks between 1989 (panel a) and 1999 (panel b) for the themes of clouds, climate, and radiation. The higher the degree of the nodes, the larger and darker (more red in the color version) they are.

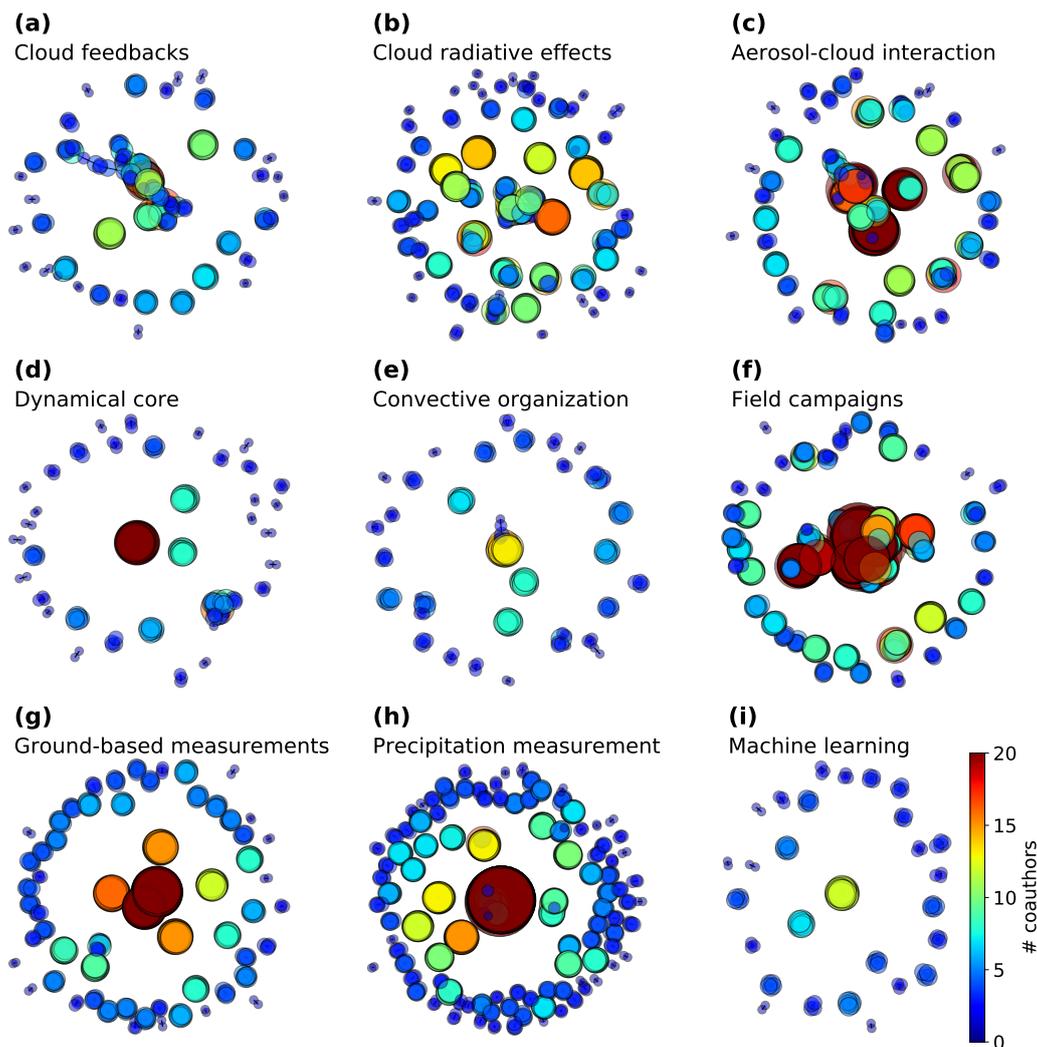


Figure 10. Work across cloud and climate research is typically characterized by a “hub” of studies with large coauthorship and a “rim” of smaller collaborations. Sample author networks for 2015 for several of the chapter themes listed in Tab. 1.

295 ture of field campaigns with many more “hub” studies than the other subfields, as well
 296 as the nascence of machine learning techniques with no large collaborations yet. Such
 297 visualizations give us an idea of where author clusters are becoming isolated and where
 298 additional community-level work or cross-institutional collaboration could be beneficial.

299 5 Summary and Outlook

300 In this introductory chapter, we have used basic SciSci analyses to analyze meta-
 301 trends in clouds and climate research. We have also motivated the layout of the over-
 302 all monograph into impacts of clouds on radiation, circulation, and precipitation. While
 303 the largest number of studies exist on clouds and radiation, recent growth in precipita-
 304 tion studies has been most rapid. Output on precipitation measurement is outpaced only
 305 by that on cloud microphysics and the recent boom in machine learning. Cloud-circulation
 306 coupling remains the least explored impact, given the large range of involved spatiotem-

307 poral scales; feasibility of large-domain storm-resolving simulations should bring new in-
 308 sight to this area. We also present three important takeaways from our analyses:

- 309 **1. As publications and interest in clouds and climate accelerate, the need**
 310 **to organize and reproduce existing work is becoming more important.**
 311 Publication on clouds and climate are doubling every 8 years with almost 2 ar-
 312 ticles published per day over the last decade. The number of different journals that
 313 publish this work is also increasing with a doubling time of 11 years. At the same
 314 time, cumulative distributions and distance metrics of citation indicate that read-
 315 ership in recent years may not be keeping pace with this burgeoning literature.
 316 As author networks expand, we are less likely to have been involved in or even heard
 317 of the efforts from others in our subfield. In the face of these trends, we need to
 318 make special effort at the individual or team level to process this information, for
 319 example using news aggregators, automatic search alerts, or journal clubs (Landhuis,
 320 2016). Regular meetings of smaller subfield communities are also crucial to pro-
 321 mote idea exchange and prevent too much “author clustering”.
- 322 **2. A synergy of techniques are fueling this growth, and we should continue**
 323 **to take advantage of complementary tools.**
 324 Only a third of abstracts on clouds and climate research can be classified as us-
 325 ing exclusively satellite data, ground-based measurement, *or* modelling. Instead,
 326 it is the combination of simulation and observation that drives the majority of re-
 327 search. We find that almost half of simulation work is still done with coarse-resolution
 328 global climate models, although high-resolution models account for another quar-
 329 ter. Increasing computational power means that cloud- and storm-resolving mod-
 330 els with kilometer- or hectometer-scale resolutions will become increasingly im-
 331 portant in the next few years (Stevens et al., 2020). A new generation of satel-
 332 lite measurements will also become available over the next years, for example with
 333 the upcoming launch of EarthCARE or the outcomes of the NASA Aerosol, Cloud,
 334 Convection, and Precipitation assessment (Illingworth et al., 2015; Gettelman et
 335 al., 2021). We emphasize the role that a synergy of models and measurements have
 336 played over the past five decades and its continued importance going forward.
- 337 **3. Small-scale cloud processes constitute the highest publication numbers,**
 338 **suggesting their intractability. Orienting the recent surge in machine**
 339 **learning toward these problems could afford progress.**
 340 Publications in microphysics, precipitation measurement, and aerosol-cloud inter-
 341 action have been the most numerous over the past ten years. These research ar-
 342 eas represent the smallest scales involved in cloud formation and evolution: droplet
 343 activation and coalescence and ice nucleation and aggregation. In the past five years,
 344 work in machine learning on clouds and climate has also exploded from 7 publi-
 345 cations in 2012 to 198 in 2019. How machine learning techniques can assist in the
 346 persistent small-scale problems of cloud physics is worthy of exploration. Such ef-
 347 forts are already underway, for example in using Bayesian inference for rain mi-
 348 crophysics or neural networks to represent aerosol effective radii and refractive in-
 349 dices (e.g., Morrison et al., 2019; Llerena et al., 2018). Neural networks have also
 350 been used to represent aspects of raindrop formation by droplet collision-coalescence
 351 and could be an even more promising route for ice microphysics, characterized by
 352 closed-form solutions or property tables rather than the coupled differential equa-
 353 tions involved in the stochastic collection equation (e.g., Gettelman et al., 2020;
 354 Seifert & Rasp, 2020).

355 While these SciSci analyses have been straightforward, they still give insight into
 356 which topics and tools are receiving the most attention. We suggest that, with its syn-
 357 thesis of scientometrics and sociology, science of science can also complement traditional
 358 review to help us understand discrepancies and biases in how clouds and climate research
 359 is done. For example, publication data mining through tools like PaperHunter (<http://>

360 paperhunter.net), Connected Papers (<https://www.connectedpapers.com>), or Schol-
 361 arSight (<http://scholarsight.org>) could be used to pinpoint contradictory studies
 362 or quantify the range in uncertain parameters across many studies. With more compre-
 363 hensive citation trees or text mining for technique identification, we would lose track of
 364 past and originating ideas less readily, avoid repeating certain studies unnecessarily, or
 365 reproduce other studies more intentionally. Such meta-level analyses would allow us to
 366 answer questions like, Are our hypotheses biased by authorship clusters? *or* Are we pro-
 367 viding sufficient funding to the intersection of subfields? Given the socioeconomic costs
 368 of climate change, clouds and climate research is done in a uniquely urgent environment,
 369 and we posit that science of science could help to optimize how this work is done in the
 370 decisive years ahead.

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 380 and search terms are given in Methods and Table 1, and the results of these searches is
 381 also directly available at <https://doi.org/10.5281/zenodo.5524758>. Codes to repro-
 382 duce figures are provided at <https://github.com/sylviasullivan/CloudLit>.

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