

A Topography of Climate Change Research

September 20, 2017

1 To contribute to evidence-based policies on tackling climate change, the IPCC aims
2 to comprehensively assess the relevant scientific literature (IPCC, 2013). With the size
3 of this literature currently almost two orders of magnitude larger than at the time of
4 the IPCC’s first assessment report, this task has become impossible without the aid of
5 machine-reading. We collect over 400,000 abstracts from Web of Science (WoS) and
6 Scopus, and develop a topic model in order to give an overview of this unmanageably
7 large corpus. This overview shows us the distribution and development of topics across
8 the literature, and allows us to identify topics with greater and lesser representation
9 in IPCC reports.

10 The size of the scientific literature on climate change has expanded rapidly over the lifetime of
11 the IPCC. While the first assessment report had around 5,500 articles to assess, nearly 5,000 new
12 articles are now published every month, bringing the total size of the literature to close to half a
13 million papers, (Figure 1). The increase in volume, velocity, and variety of content to be assessed
14 has turned the task of the IPCC into a ‘Big Literature’ challenge (Minx, 2017). To ask questions
15 about the literature *at scale*, we now need to apply computational techniques to the analysis of large
16 document collections.

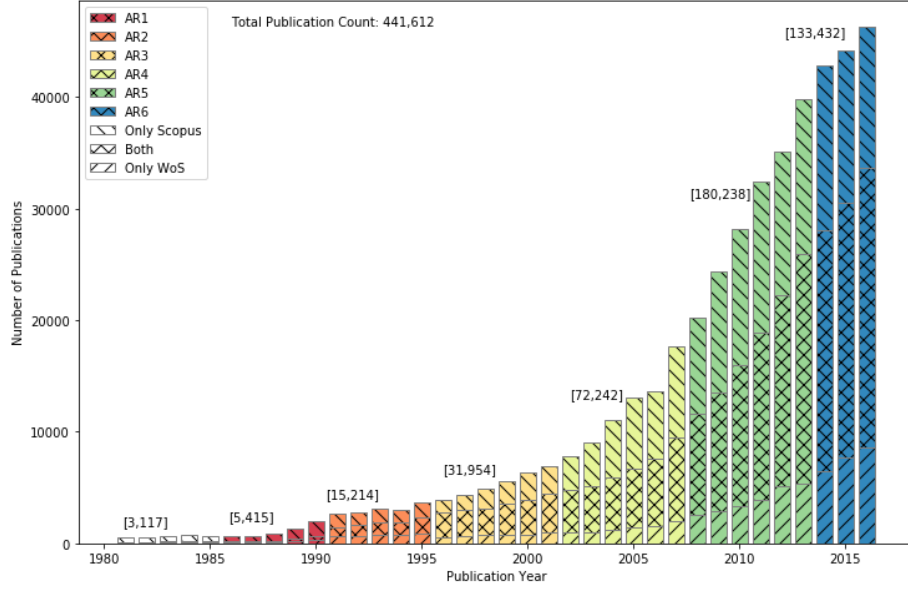


Figure 1: Growth in relevant literature in WoS and Scopus

Topic models are one such technique. A topic model learns the latent topics that structure a large corpus of documents, by leveraging the systematic co-occurrence of words across documents. Topics are distributions of words, and the topic mixture of each document explains the words observed in that document. This means that topic models can aid the understanding of large corpora, and of the place of individual documents within them, by showing a document or corpus as a combination of 100 or so intelligible topics, rather than combinations of thousands of words.

Assessment-makers like the IPCC have been described as cartographers for policymakers (Edenhofer and Kowarsch, 2015). As such their purpose is, summarising available scientific knowledge, to describe the problem and solution space of a policy issue. The topic model presented here is a rough map of climate change research since 1985. It shows a broad outline of the topics that make up this research and how they relate to each other, and demonstrates how this has changed over time. Such a map sheds light on the terrain of knowledge about a policy issue, making an

29 overview of an unmanageably large and diverse landscape possible. This overview allows both for
30 the production of policy pathways that are well informed by science, and, as demonstrated in this
31 paper, the assessment of the comprehensiveness with which these pathways reflect the landscape.

32 While topic modelling has been employed to answer specific questions about small aspects of
33 climate literature, e.g. (e.g. Minx et al., 2017; Grubert and Siders, 2016), this is the first application
34 of topic models to gain an overview of the entire field.

35 Results

36 37 Topic Model

- 38 • Figure 2 shows the structure of the topic model with 100 topics, with each node coloured
39 according to the IPCC working group in which the highest proportion of the topic's documents
40 are cited.
- 41 • Topics that systematically co-occur in documents are linked, and the resulting network is
42 displayed with links weighted according to topic correlation score. The strength of links is
43 greater where two topics are categorised as being in the same working group. This relationship
44 is statistically significant (see SI)
- 45 • Some big/interesting topics are x and y
- 46 • These topics have grown at particularly interesting times (Figure 3)
- 47 • These topics are better covered in IPCC, these are less well covered 4

From here
on in, the
paper is
more of an
outline, with
a mixture of
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about results
that exist
or I would
still like to
generate

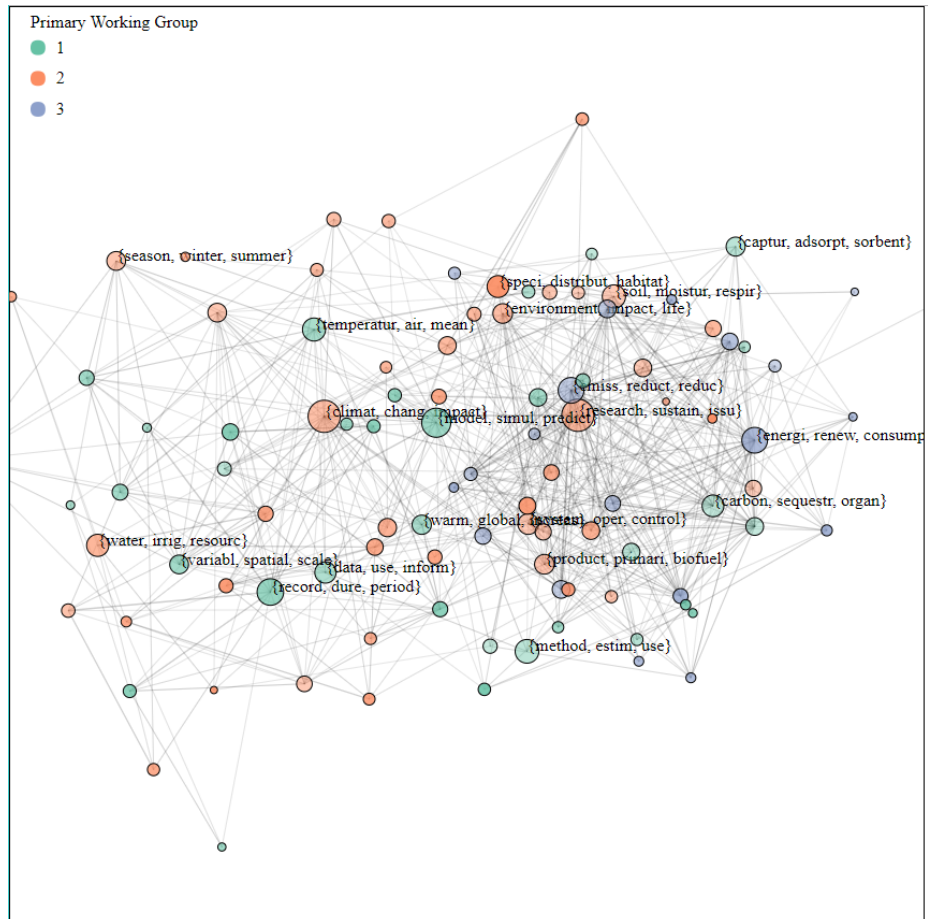


Figure 2: Topic structure of climate change literature

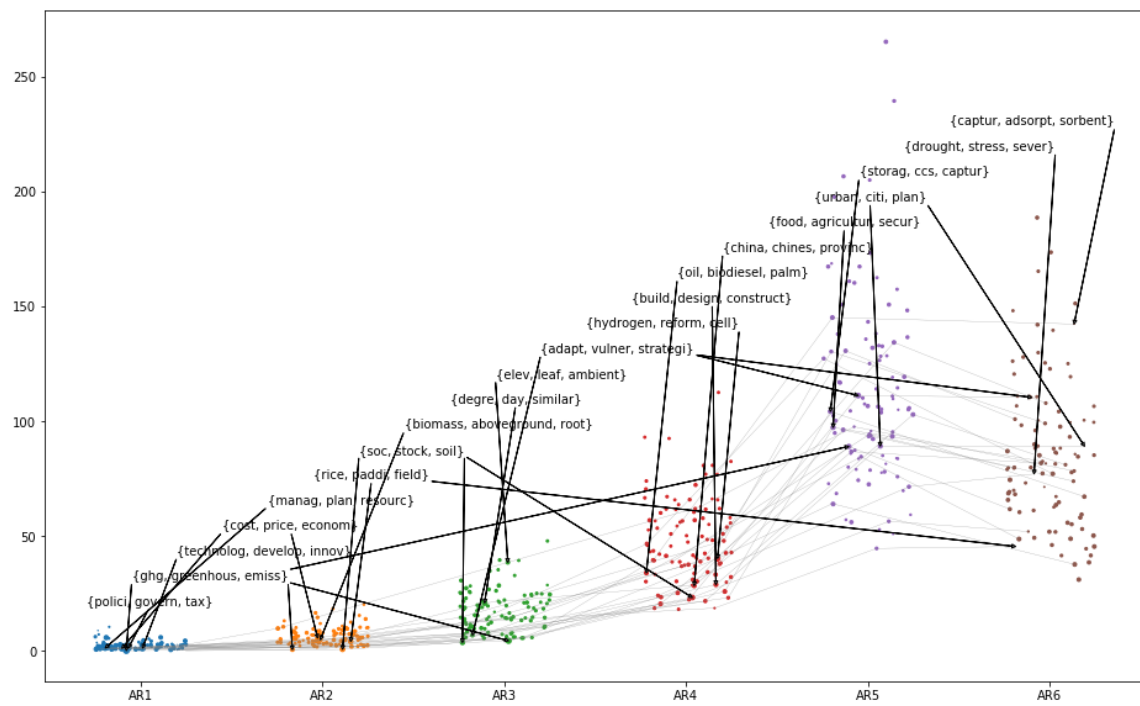


Figure 3: Topic growth over time. The 3 topics in each assessment period that grew by the largest amount are labelled [Example figure]

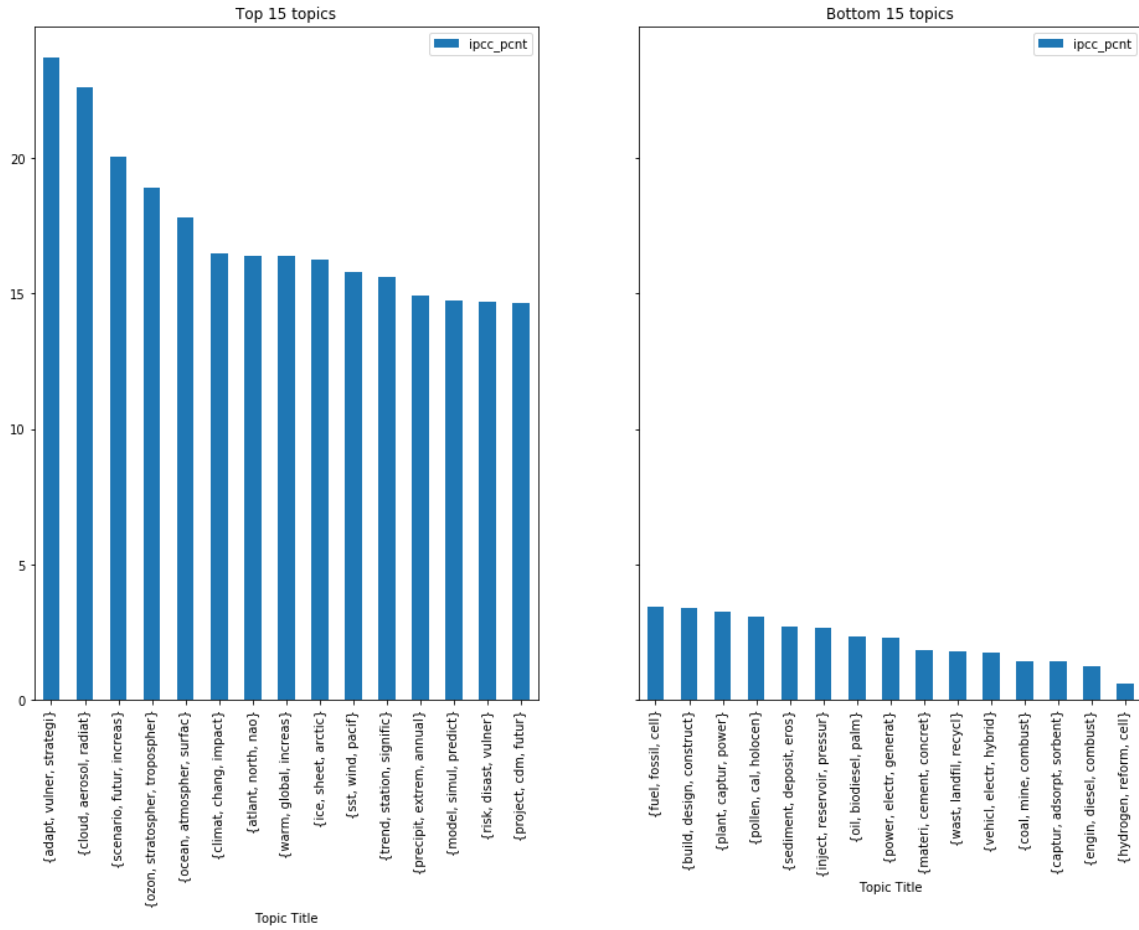


Figure 4: IPCC references by topic. The bars show the percentage of each topic that has been matched to an IPCC reference

AR1		AR2		AR3		AR4		AR5		AR6	
title	pchange	title	pchange	title	pchange	title	pchange	title	pchange	title	pchange
polici,	255%	rice,	513%	ghg,	639%	hydrogen,	618%	adapt,	397%	drought,	8%
govern,		paddi,		green-		reform,		vulner,		stress,	
tax		field		hous,		cell		strategi		sever	
ghg,	250%	soc,	492%	emiss		soc,	534%	ghg,	334%	rice,	2%
green-		stock,		degre,	566%	stock,		green-		paddi,	
hous,		soil		day,		soil		hous,		field	
emiss		cost,	398%	similar		build,	417%	emiss		urban,	-0%
technolog,	240%	price,		soc,	432%	design,		food,	320%	citi,	
develop,		econom		stock,		con-		agri-		plan	
innov		biomass,	387%	soil		struct		cultur,		adapt,	-1%
cost,	220%	above-		elev,	294%	oil,	402%	secur		vulner,	
price,		ground,		leaf,		biodiesel,		urban,	311%	strategi	
econom		root		ambient		palm		citi,		captur,	-2%
manag,	191%	ghg,	354%	adapt,	278%	china,	384%	plan		adsorpt,	
plan,		green-		vulner,		chines,		storag,	303%	sorbent	
resourc		hous,		strategi		provinc		ccs,			
		emiss						captur			

Table 1: The top 5 topics in each assessment period by percentage growth since the last assessment period

AR1		AR2		AR3		AR4		AR5		AR6	
title pchange		title pchange		title pchange		title pchange		title pchange		title pchange	
tree,	-11%	sediment,	91%	ice,	101%	concentr,	87%	cloud,	47%		
stand,		deposit,		sheet,		atmo-		aerosol,		technolog,	-39%
ring		eros		arctic		spher,		radiat		develop,	
soc,	-15%	drought,	84%	food,	100%	air		ice,	45%	innov	
stock,		stress,		agri-		methan,	86%	sheet,		coal,	-40%
soil		sever		cultur,		oxid,		arctic		mine,	
coal,	-19%	sea,	84%	secur		wetland		sst,	42%	combust	
mine,		level,		coal,	75%	elev,	63%	wind,		fuel, fos-	-40%
combust		coastal		mine,		leaf,		pacif		sil, cell	
dioxid,	-23%	lake,	83%	combust		ambient		atlant,	35%	hydrogen,	-41%
carbon,		level,		warm,	73%	rice,	49%	north,		reform,	
atmo-		diatom		global,		paddi,		nao		cell	
spher		glacier,	61%	increas		field		ozon,	25%	oil,	-42%
solar,	-31%	mass,		dioxid,	67%	degre,	37%	strato-		biodiesel,	
radiat,		retreat		carbon,		day,		spher,		palm	
irradi				atmo-		similar		tropo-			
				spher				spher			

Table 2: The bottom 5 topics in each assessment period by percentage growth since the last assessment period

Figure 5: Focus on [biochar?] showing document with highlighted words

Figure 6: Model validation graph, showing error for different topic numbers, feature numbers

- Some interesting topic correlations are x and y; well fitting documents to both include x and y
-

Conclusion

- A very simple topic model provides an overview of the whole landscape.
- This allows researchers / assessment makers to identify areas that have grown recently
- Topic models aid document discovery, have the potential to contribute to more comprehensive assessments.
- AR5 seemed to have less comprehensive coverage of x topics. This was a particular issue in WG y.
- This may not be an issue, there could be good reasons for this, but these should be made transparent.
- For the next assessment report, x topics may require particular attention.
- The emerging topic on CCS resonates with a growing recognition of the importance of negative emissions and the lack of understanding about how they could fill their role. This will be of particular importance for the IPCC special report on 1.5 degrees.

Figure 7: Some relation of topics to other features of dataset: e.g. most interdisciplinary journals and least, or so...

64 Methodology

- 65 • Topic modelling in general: reducing large matrix of documents to words to two smaller
- 66 matrices of topics x words and topics x documents.
- 67 • Model selection: NMF (Lee and Seung, 1999)
- 68 • How does it work? Advantages: Simple, scalable: better results than with other solutions
- 69 • Topic model browser Chaney and Blei (2012)
- 70 • Merging with IPCC citation dataset - caveats...
- 71 • Network explanation
- 72 • Regression of network score on dummy variable for same

73 1 Data

- 74 • Queries: use Grieneisen and Zhang (2011), or take the best bits of Grieneisen and Zhang (2011)
- 75 and Haunschild et al. (2016)?
- 76 • Sources: WoS, Scopus or both?
- 77 • Preprocessing: Remove punctuation, numbers, common, uncommon words, stemming

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