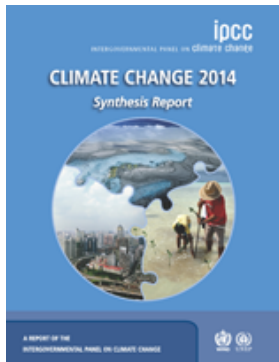


A Topography of Climate Change Research

Max Callaghan

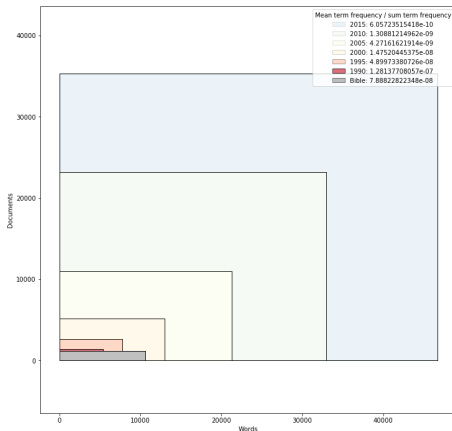


February 5, 2018



- To contribute evidence-based policy-making on climate change, the IPCC aims to *comprehensively* assess
- These assessments should be aim to balance legitimacy, credibility and relevance (?)

These extra 36,000 papers take the total number of documents to 309,697.



- Topic modelling is a way of reducing the dimensionality of a corpus of documents
- A large matrix of documents \times words is factorised by a matrix of topics \times words and a matrix of topics \times documents (?)
- Topics describe the latent structure of the document corpus

$$V_{i\mu}$$

V: 8769 x 3495



Figure: A topic model of 3495 documents on climate change from the year 2000

$$V_{i\mu} \approx (WH)_{i\mu} = \sum_{a=1}^r W_{ia} H_{a\mu}$$

V: 8769 x 3495

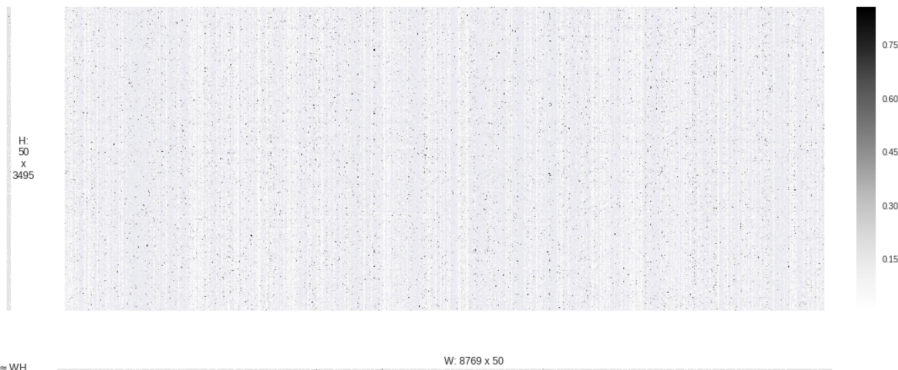


Figure: A topic model of 3495 documents on climate change from the year 2000

- What is the thematic structure of the literature on climate change, and how has this changed over the five assessment periods of the IPCC
- What can this modelled thematic structure tell us about the past and future relationship between the IPCC and scientific literature on climate change?

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- Dynamic Topic Modelling (DTM) (?) assume that a constant number of topics exists over all topic models, but allows the words in the topics to evolve from one time period to another
- Dynamic Non-negative Matrix Factorisation (?) has varying numbers of topics in each window and allows for topics to emerge and/or disappear.

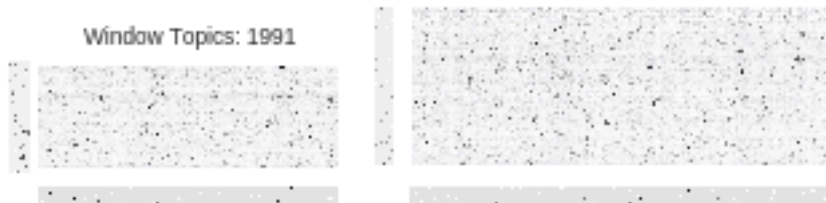
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Where the size and variety of the literature we want to model has increased exponentially, we need an approach that allows for the emergence of new topics.

Window Topics: 1991





Window Topics: 1992

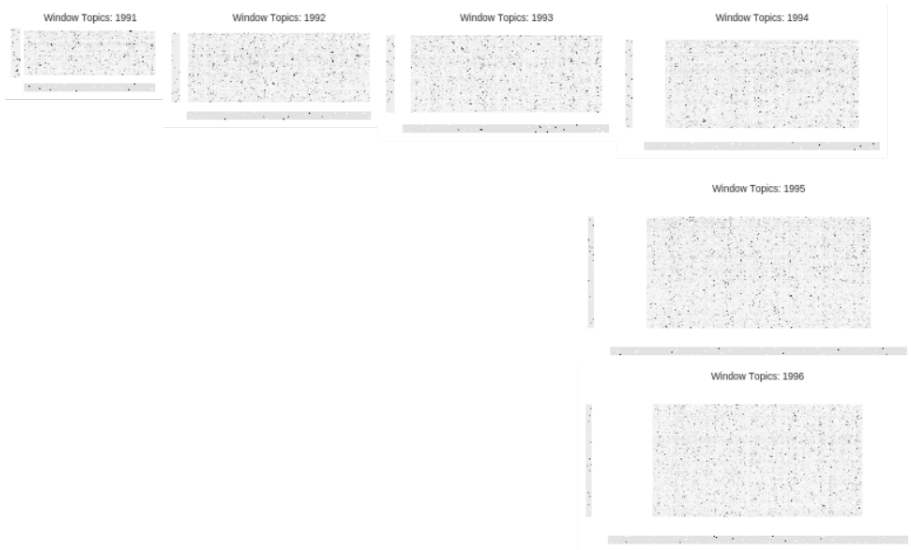


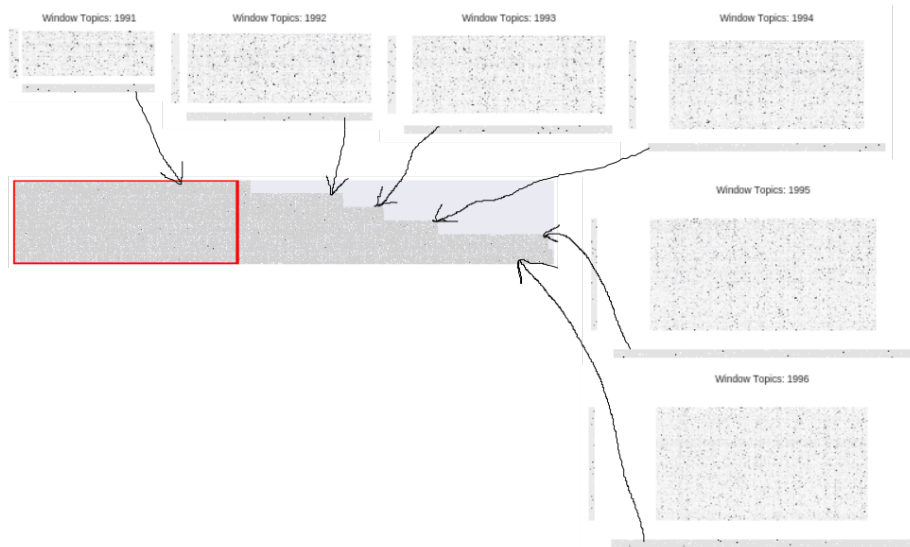
Window Topics: 1991

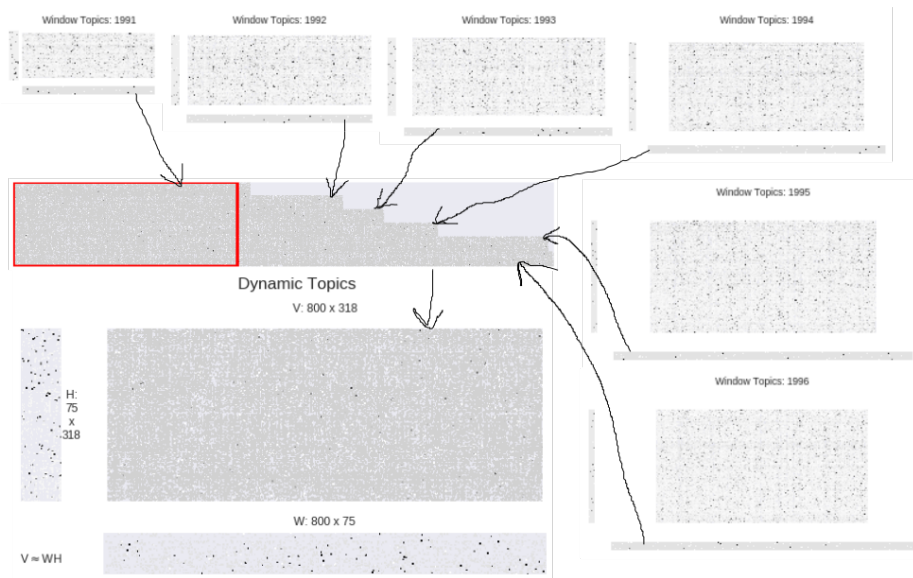


Window Topics: 1993



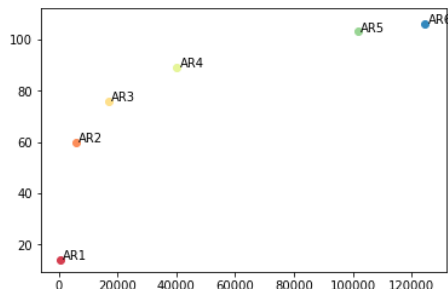






- Choosing the number of window topics is non-trivial. Data-driven approaches are limited (see below), and human selection is time consuming.
- To facilitate the description of trends over the assessment periods of the IPCC, and to minimize the number of modelling decisions, I consider each IPCC assessment period as a time window.

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- To facilitate the description of trends over the assessment periods of the IPCC, and to minimize the number of modelling decisions, I consider each IPCC assessment period as a time window.
- Starting from a logarithmic relationship between the number of documents and the ideal topic number, I compare 5 runs with varying numbers of topics for each window



Human topic number criteria

- Intelligibility

Data-driven topic number criteria

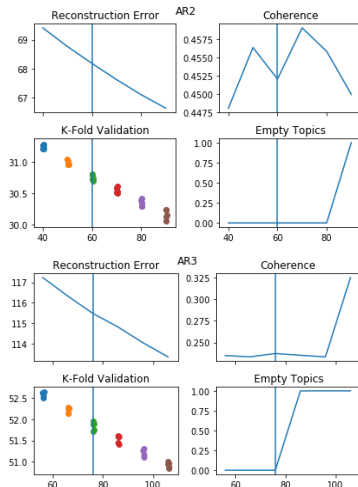
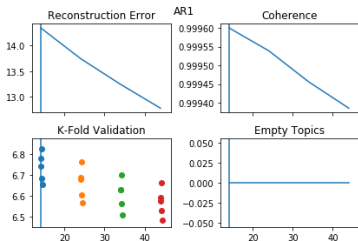
- Reconstruction accuracy
- Predictive capacity

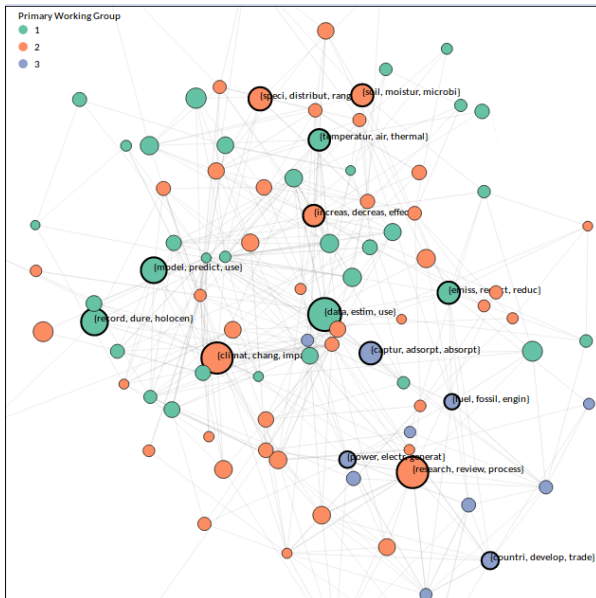
Human topic number criteria

- Intelligibility

Data-driven topic number criteria

- Reconstruction accuracy
- Predictive capacity





- A network of comprehensible topics is generated with 100 topics
- Topics can be matched to the IPCC working group from which the majority of the topic documents are referenced in
- Topics from the same working group are **significantly** more likely to be correlated with each other than those which are not

IPCC Coverage	Primary WG	Topic Title	WG 1	WG 2	WG 3
0.16%	1	{rainfal, monsoon, rain}	0.50%	0.50%	0.00%
0.10%	2	{veget, ndvi, cover}	0.41%	0.59%	0.00%
0.16%	1	{snow, cover, winter}	0.59%	0.41%	0.00%
0.17%	2	{region, local, scale}	0.41%	0.59%	0.00%
0.16%	1	{coastal, mangrov, rise}	0.57%	0.42%	0.01%

IPCC Coverage	Primary WG	Topic Title	WG 1	WG 2	WG 3
0.09%	3	{gas, coal, greenhous}	0.30%	0.15%	0.56%
0.10%	3	{transport, vehicl, road}	0.24%	0.12%	0.64%
0.13%	1	{emiss, reduct, reduc}	0.45%	0.21%	0.34%
0.09%	1	{methan, oxid, methanotroph}	0.63%	0.16%	0.20%
0.13%	3	{ghg, greenhous, gas}	0.15%	0.09%	0.75%

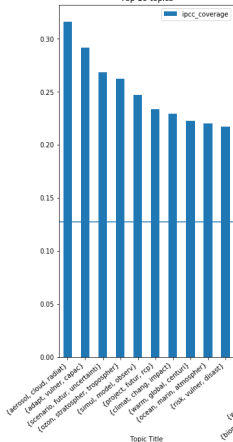
IPCC Coverage	Primary WG	Topic Title	WG 1	WG 2	WG 3
0.11%	2	{sustain, develop, resourc}	0.04%	0.51%	0.46%
0.08%	3	{build, construct, design}	0.03%	0.38%	0.59%
0.11%	2	{environment, impact, life}	0.06%	0.58%	0.36%
0.19%	3	{polici, tax, govern}	0.02%	0.32%	0.66%
0.16%	2	{urban, citi, plan}	0.07%	0.55%	0.38%

How can we get a sense of which topics are better covered in IPCC reports?

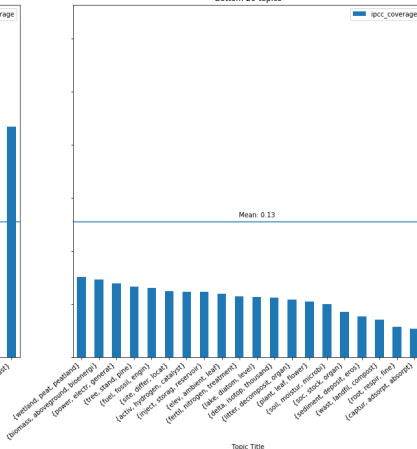
- Each document d either matches or does not match an IPCC reference
- For each topic h , we can sum the scores for each category of document
- The “IPCC proportion” of each topic is the proportion of the sum of the document score accounted for by documents which match IPCC references.

Preliminary results - gaps in coverage

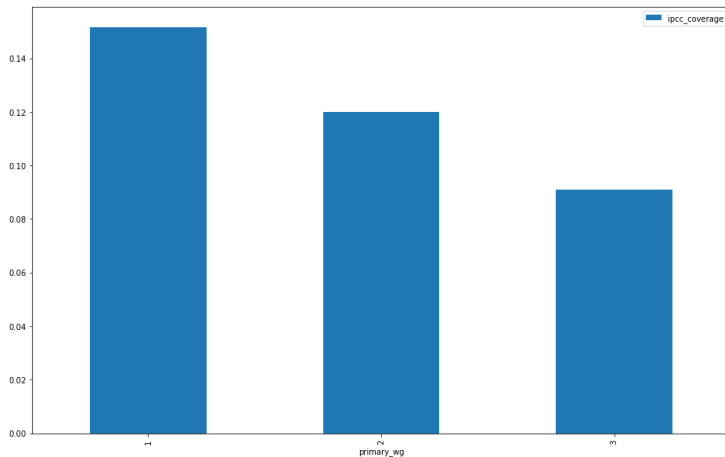
Top 10 topics



Bottom 20 topics



- The physical science aspects of climate change, as well topics on impacts, adaptation and scenarios are well covered by the IPCC
- Topics on specific technological solutions (particularly NETs), as well as soils, are less well covered



- Adjusting second stage of model to give more weight to topics with many documents (to better capture emergence)
- Further analysis of new model
- (Manual) comparison of (AR6) topics with AR6 outline

