

# A Rapid Computer-assisted Systematic Map of Regional Climate Impacts - Results [1]



May 14, 2020

# Outline

Recap - Goal

Data collection

Outcome 1 - prediction performance

Outcome 2 - Evidence Map

## Goal

There are hundreds of thousands of documents potentially relevant to observed climate impacts. We want to be able to do two things:

- ▶ Separate those documents which *are* relevant from those that are not
- ▶ Predict in what way relevant documents are relevant:
  - ▶ What impacts do they document?
  - ▶ What type of evidence do they provide?
  - ▶ In which locations is there evidence

Once we can do that, we can draw a rough map of the available evidence, and/or aid the production of an *assessment* of the available evidence

# Context

Systematic assessments of the evidence on Climate Change like those conducted by the IPCC are vital.

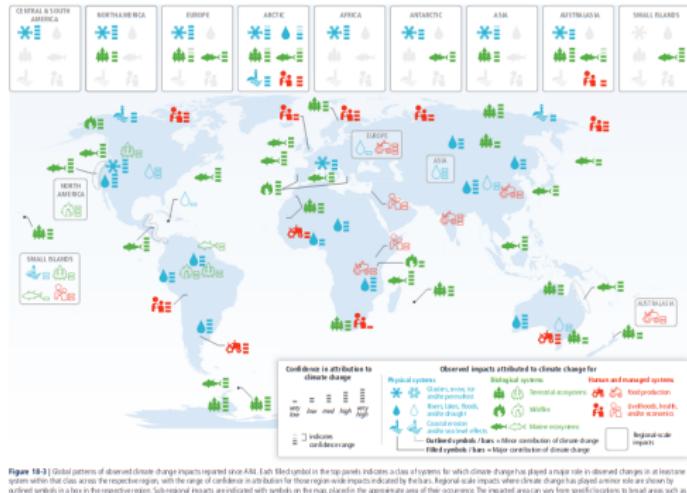


Figure 18-3 | Global patterns of observed climate change impacts reported to AR5. Each symbol in the map indicates the range of confidence for which climate change has played a major role in observed change in that area. Symbols that also appear in the maps for regions with the range of confidence as contribution for those regions are repeated in the figure. Regional-scale impacts are shown as outlined symbols in a box in the corresponding region. Sub-regional impacts are indicated with symbols on the map, placing the approximate area of their occurrence. The impacted areas can vary from specific locations to broad areas such as a major river basin. Impacts on physical (blue), biological (green), and human (red) systems are differentiated by color. This map represents a graphical synthesis of Tables 18-5, 18-6, 18-7, 18-8, and 18-9. Absence of climate impacts from this figure does not mean that such impacts have not occurred.

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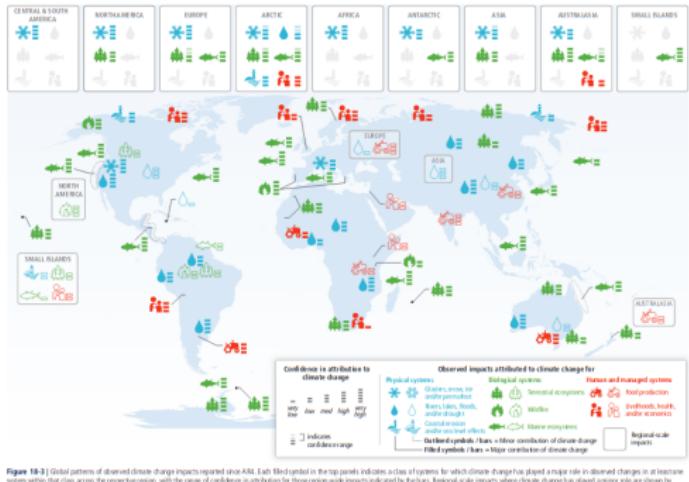
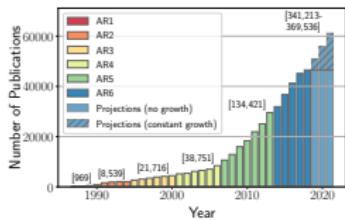


Figure 18-3 | Global patterns of observed climate change impacts mapped by AR. Each block relates to the range of confidence in the range of confidence in attribution of climate change to observed changes in at least one system that also arises from the impacts being mapped. Sub-regional impacts are indicated by symbols on the map, placing the approximate area of their occurrence. The impacted areas can vary from specific locations to broad areas, such as a major river basin. Impacts on physical (blue), biological (green), and human (red) systems are differentiated by color. This map represents a graphical synthesis of Tables 18-5, 18-6, 18-7, 18-8, and 18-9. Absence of climate change impacts from this figure does not mean that such impacts have not occurred.



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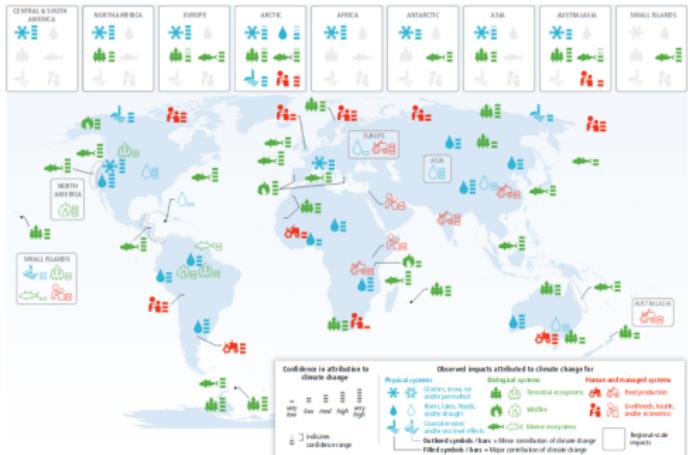
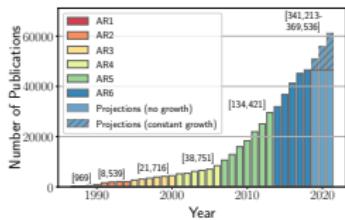


Figure 18-3 | Global patterns of observed climate change impacts reported in AR4. Each colored box spans to the top-right indicates the range of confidence in attribution for each region-wide impacts indicated by symbols. Regional-scale impacts are indicated by a different symbol in a box in the respective region. Sub-regional impacts are indicated with symbols on the map, placing the approximate area of their occurrence. The impacted areas can vary from specific locations to broad areas such as a major river basin. Impacts on physical (blue), biological (green), and human (red) systems are differentiated by color. This map represents a graphical synthesis of Tables 18-5, 18-6, 18-7, 18-8, and 18-9. Absence of climate change impacts from this figure does not mean that such impacts have not occurred.



- ▶ These are challenged by big literature
- ▶ They do not account for uncertainty about what literature is available

## Distribution of labour between humans and machines

A human expert or a team of human experts is best placed to answer those questions for any single document, but they can't look at all potentially relevant documents

We can use labels generated by humans to try to teach a computer what a relevant document looks like, and how to decide in what way it is relevant.

If this works well, we can predict, with some uncertainty, how much evidence there is, and where and on what topic it is.

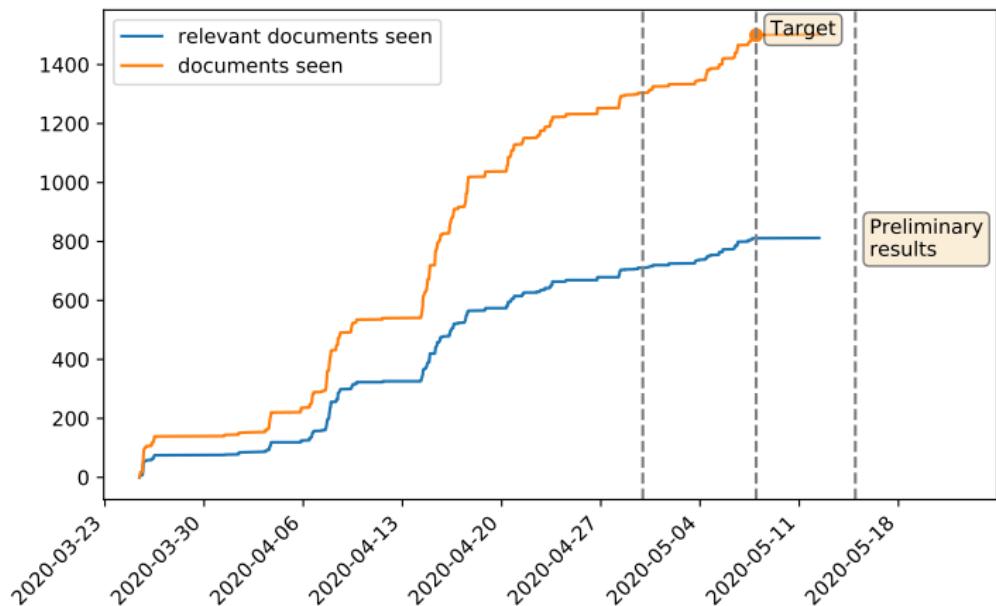
Recap - Goal

## Data collection

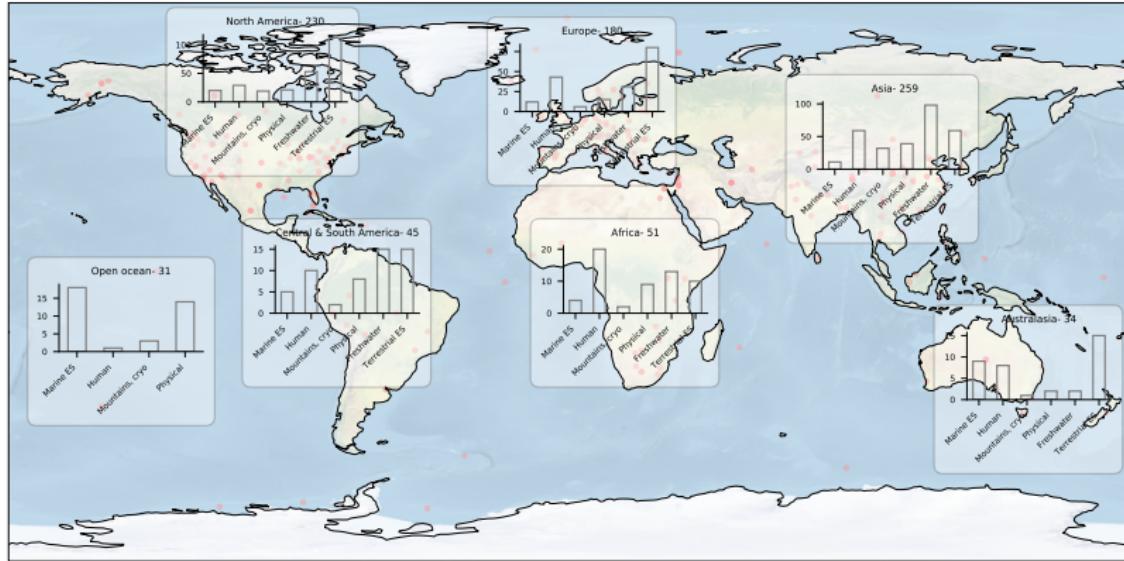
Outcome 1 - prediction performance

Outcome 2 - Evidence Map

Over the last month and a half we screened 1500 documents



# This already constitutes a useful information gathering exercise



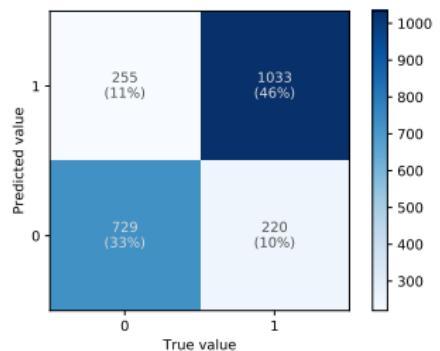
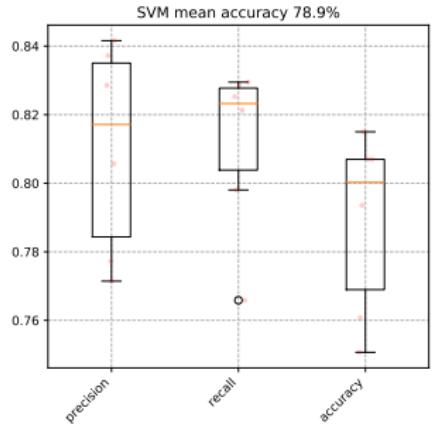
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Data collection

Outcome 1 - prediction performance

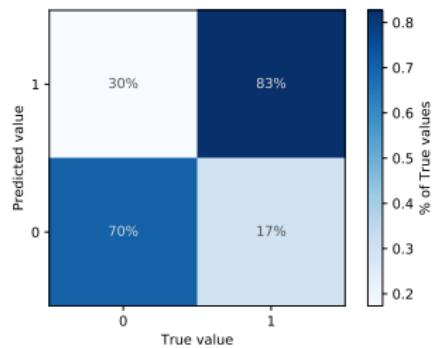
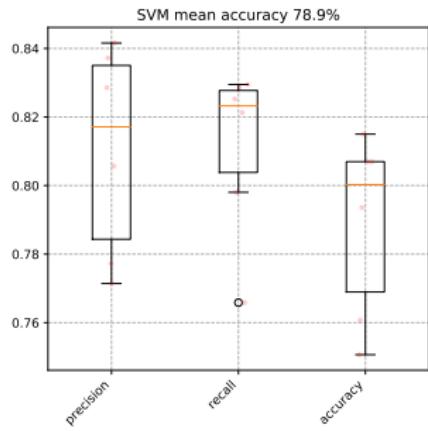
Outcome 2 - Evidence Map

## We predict the relevance of a document most of the time



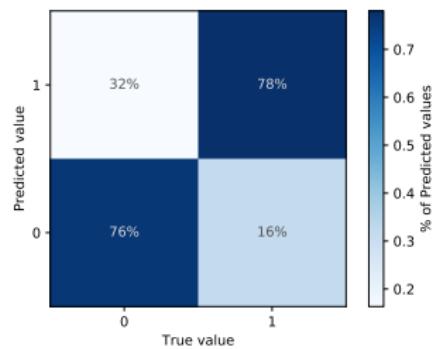
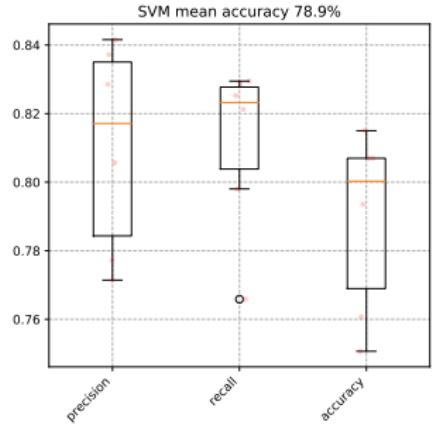
This has been steadily increasing by using the model itself as a "second pair of eyes" to check for errors, and I expect it to increase further (partly due to different criteria for inclusion at different stages of the project)

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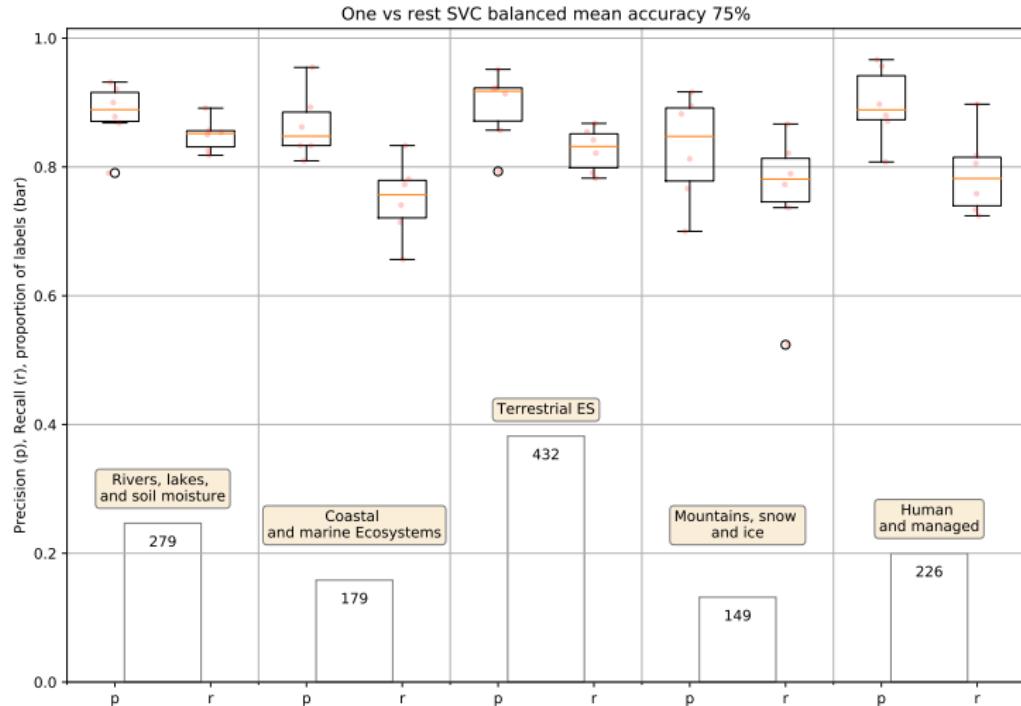
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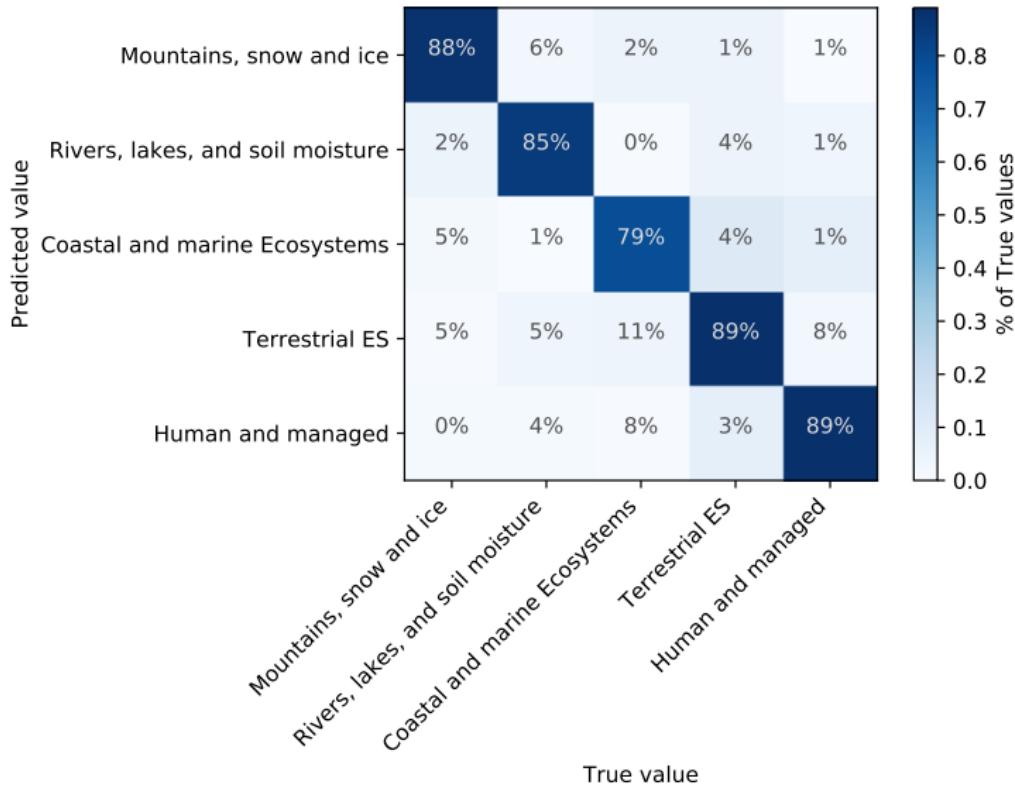
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We can clearly identify what impact category a document is related to

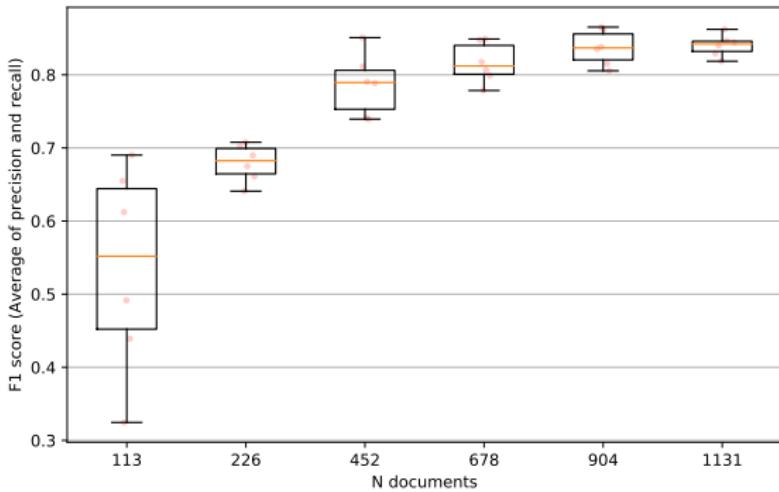


**Figure:** Precision (how many documents predicted to be in a category actually had that label) and Recall (how many documents with a label were predicted to be in that category) for each broad impact category

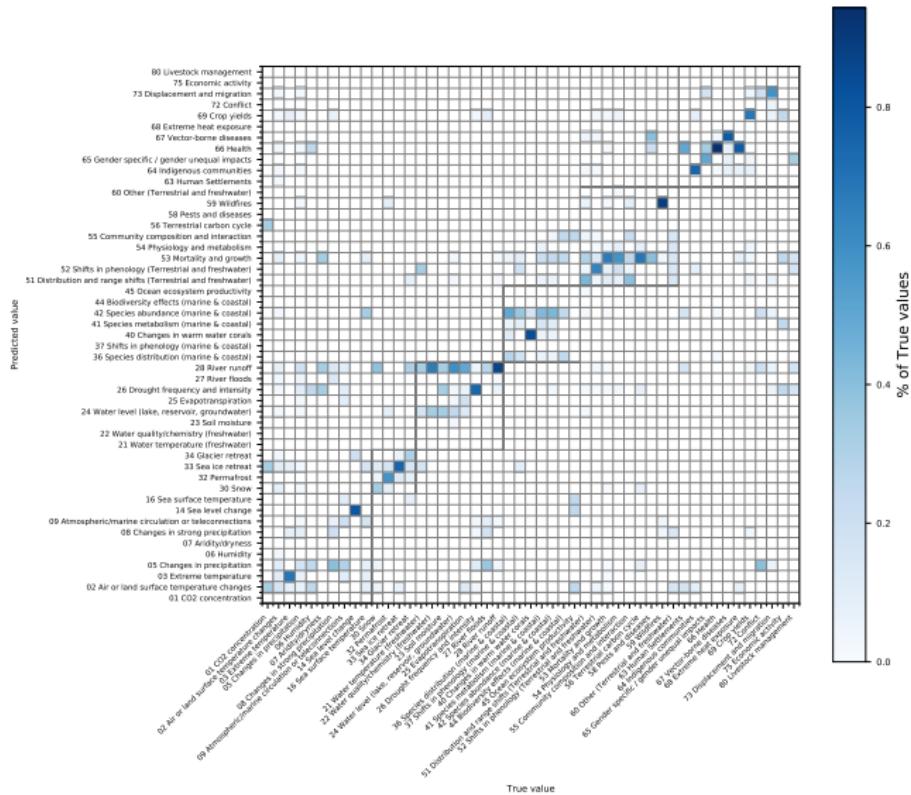
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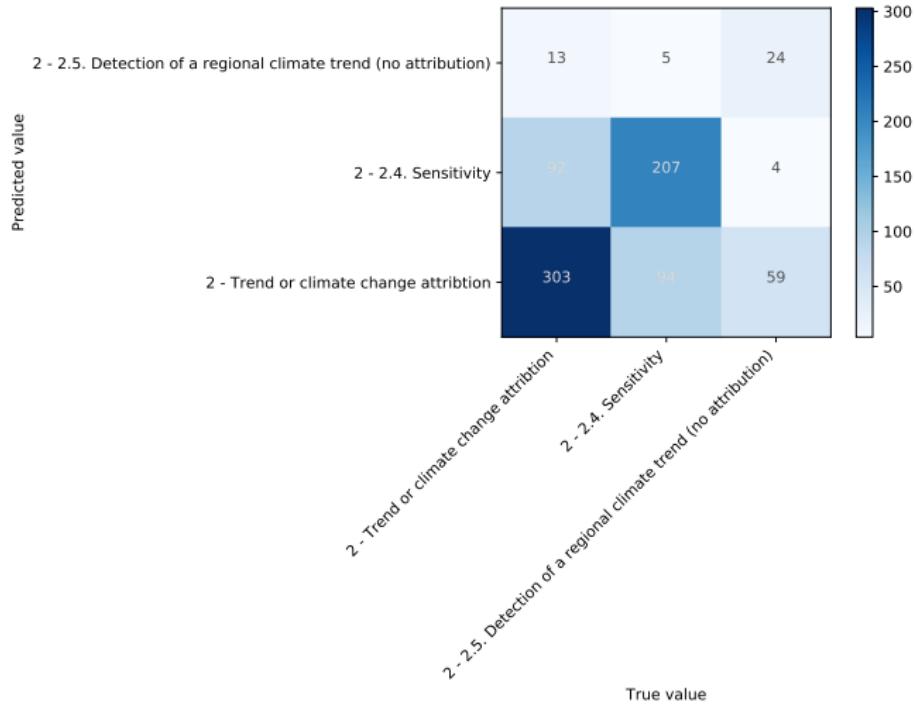
## Accuracy and uncertainty increased with the number of labels available



We are also broadly correct on subcategories - impressive given the amount of data and the complexity of the coding scheme



## Getting attribution correct is harder, but we have fewer labels



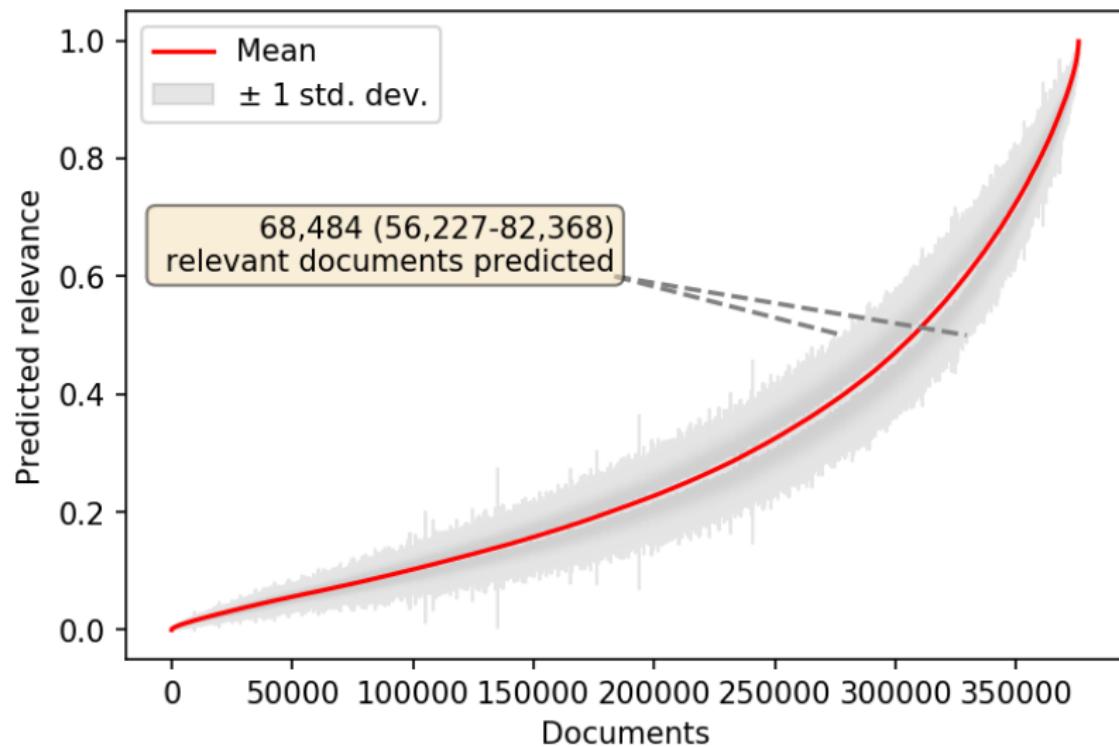
Recap - Goal

Data collection

Outcome 1 - prediction performance

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We predict tens of thousands of additional documents relevant according to the criteria we defined

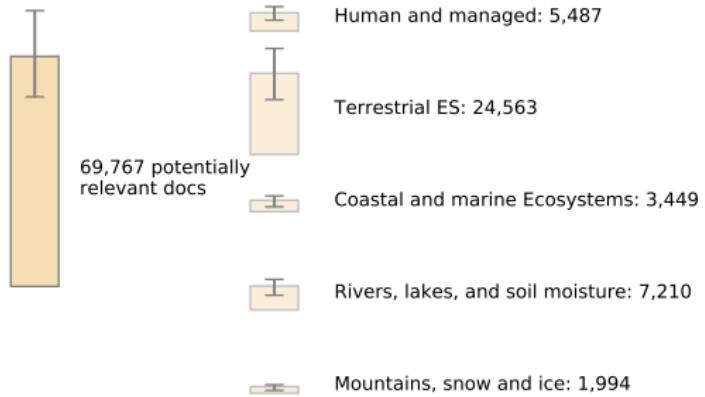


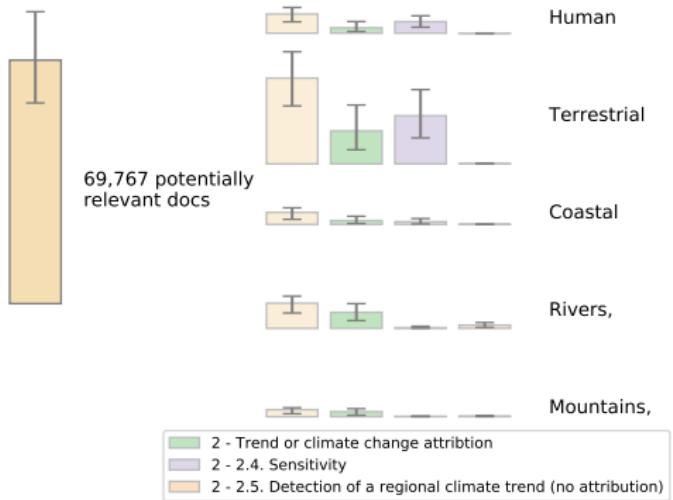


378,365 potentially  
relevant docs

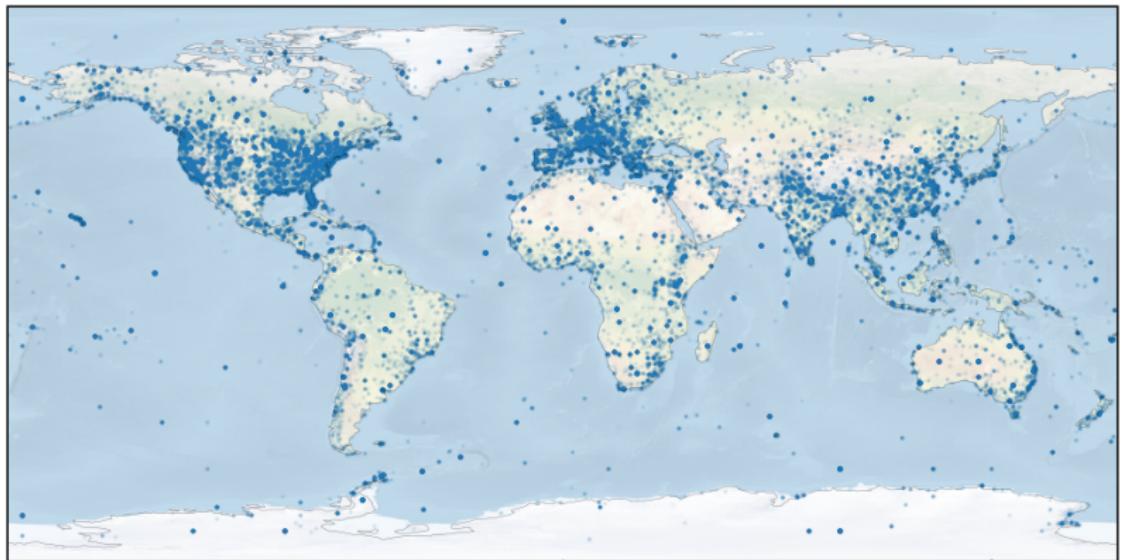


69,767 relevant  
documents



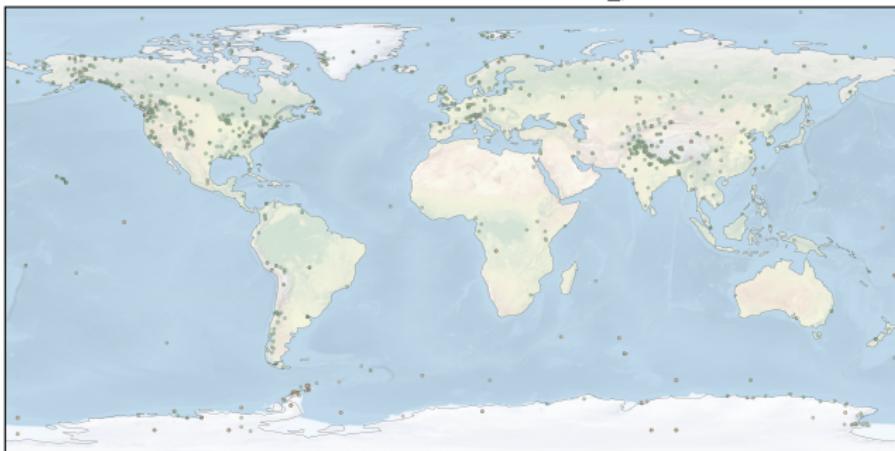


The studies predicted to be relevant cover a much broader array of places, but geographic imbalances persist



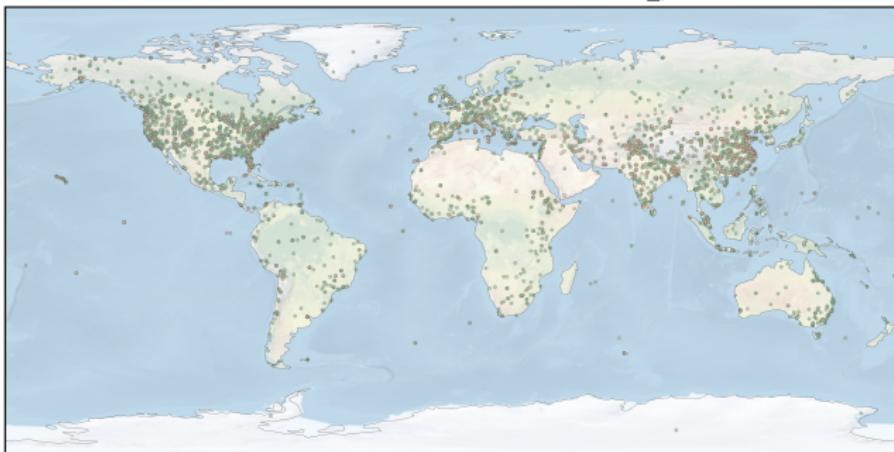
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12 - Mountains, snow and ice - mean\_prediction



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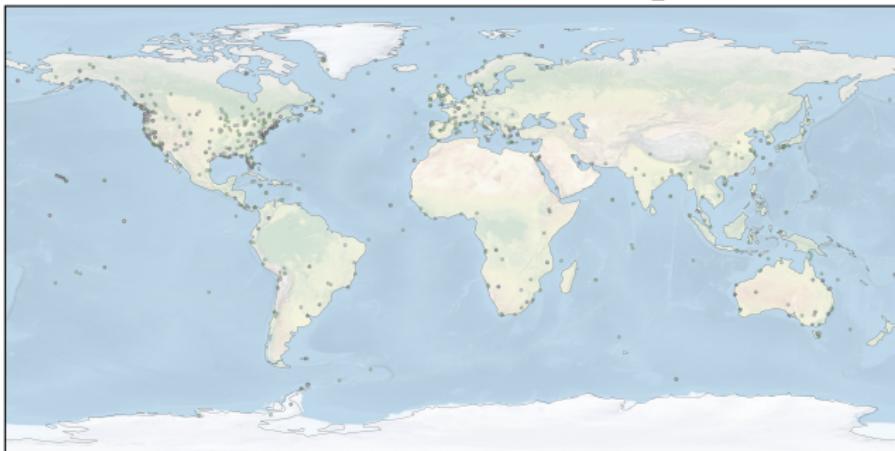
12 - Rivers, lakes, and soil moisture - mean\_prediction



- 2 - Trend or climate change attribution - mean\_prediction
- 2 - 2.5. Detection of a regional climate trend (no attribution) - mean\_prediction
- 2 - 2.4. Sensitivity - mean\_prediction

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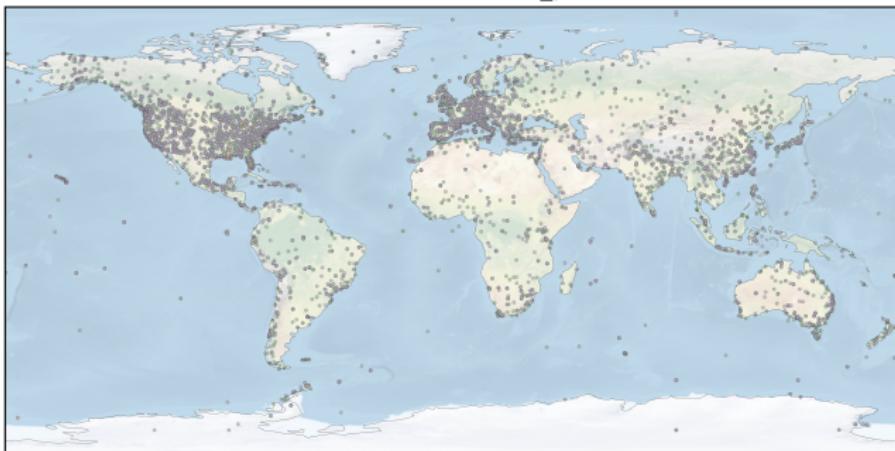
12 - Coastal and marine Ecosystems - mean\_prediction



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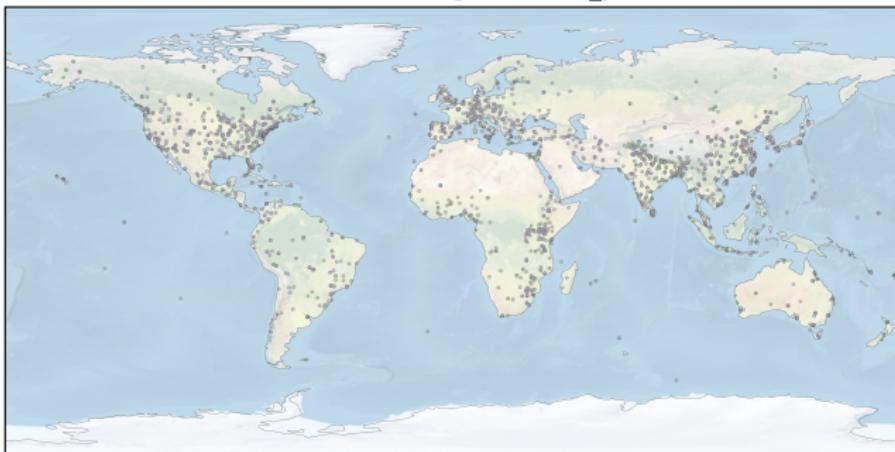
12 - Terrestrial ES - mean\_prediction



- 2 - Trend or climate change attribution - mean\_prediction
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- 2 - 2.4. Sensitivity - mean\_prediction

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12 - Human and managed - mean\_prediction



- 2 - Trend or climate change attribution - mean\_prediction
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- 2 - 2.4. Sensitivity - mean\_prediction

# Outlook

## So far

- ▶ Data collection
- ▶ Coding scheme
- ▶ Coding
- ▶ Learning and predictions
- ▶ Collation of results

## Still to come

- ▶ Further data checking
- ▶ Investigating distribution of evidence and comparing with IPCC
- ▶ Predicting drivers and mapping driver-impact pathways
- ▶ Write up
- ▶ Interactive map