

11-12 NOVEMBER 2021 PARIS

13h30-14h30

RESPONSIBLE AI WORKING GROUP

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Nicolas Miailhe
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THE RESPONSIBLE AI WORKING GROUP'S MANDATE

The mandate of the Working Group is "foster and contribute to the responsible development, use and governance of human-centred AI systems, in congruence with the UN Sustainable Development Goals".

CLIMATE CHANGE AND AI

Recommendations for Government Action







Raja Chatila
Co-Lead
Sorbonne University



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Co-Lead
Founder & President
The Future Society



RESPONSIBLE AI FOR SOCIAL MEDIA GOVERNANCE

A proposed collaborative method for studying the effects of social media recommender systems on users



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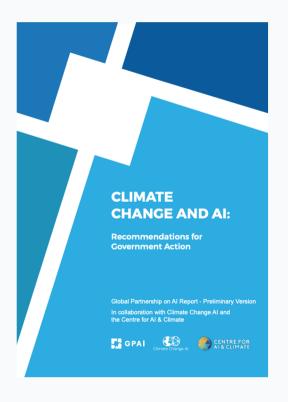
Dino Pedreschi
Co-Lead
Professor of Computer Science
University of Pisa

Our new Co-Chair: 2022-24



Catherine Régis
Co-chair Responsible Al
Canada Research Chair in
Health Law and Policy
Professor of Law
Université de Montréal

A Responsible Al Strategy for the Environment







Responsible Al Working Group

Nicolas Miailhe, Committee Co-Lead Raja Chatila, Committee Co-Lead Marta Kwiatkowska, Committee Steering Group

Overall Objective

"Develop a global responsible Al adoption strategy for climate action and biodiversity"

Short-term Objectives

"Create a roadmap of AI & Climate action ahead of the COP-26"



- Analyse responsible Al benefits & risks
- Build **roadmaps** for govs, IGOs, and research
- Develop catalogue of **high-impact Al use cases**

Long-term Objectives



Strengthen and expand the climate action roadmap



Work with **institutional partners** to anchor the climate action roadmap at the **COP** and other forums **ECMWF**













Expand the scope to include biodiversity promotion in other GPAI projects



Develop an **impact and risk assessment framework** harnessing
Al for climate action and biodiversity
preservation responsibly





Climate Change and Al Recommendations for Government Action





Report developed in collaboration between members of Climate Change AI and the Centre for AI & Climate, and experts in the Global Partnership on Artificial Intelligence's Committee on Climate Action and Biodiversity Preservation, as part of the broader working group on Responsible AI. The report reflects the personal opinions of the authors and does not necessarily reflect the views of the experts' organizations, GPAI, the OECD, or their respective Members.

Report Structure & Areas of Impact



Al as a tool for climate action



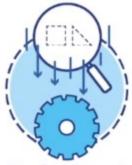
Data & digital infrastructure

Data, simulation environments, testbeds, libraries, computational hardware



Research & innovation funding

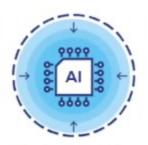
Interdisciplinary & cross-sectoral work guided by climate impact



Deployment & systems integration

Policy design & evaluation, market design, business models





Reducing Al's negative impacts on the climate

Application and compute-related impacts









Responsible Al



Capacity building



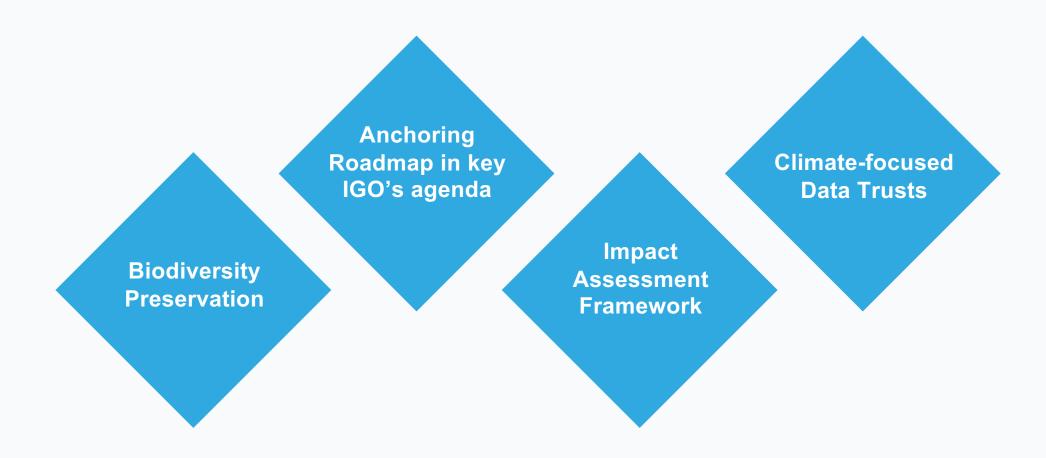
International collaboration



Impact assessment



Work Plan 2022





Responsible Al for Social Media Governance





Three related sub-projects

COMMUNITY CONSULTATION

Asking the citizens of a given country (NZ) more generally what counts as 'harmful content'

TECHNICAL

Working with social media companies to investigate whether their recommender systems move users towards 'harmful content'

LEGAL/POLICY

What's the law/policy basis for the proposed engagement with social media companies?



Focus of technical project

Rec Sys

Focus is on the recommender systems that deliver content into social media users' 'feeds'.

Learning

RecSys are Al/ML systems that learn about what each user likes to engage with. Through learning, RecSys deliver content that's personalised to each user.



Scientists have concerns about how Recommendation Systems learn

- Recommendation Systems learn from seeing which feed items a user clicks on (or otherwise engages with)
 - user clicks → recsys learning
- But the user chooses from a list of items the Rec Sys already thinks she will like
 - o recsys learning → user clicks
- There's a feedback loop here, which can lead to instabilities.
- Users also show certain systematic biases in their clicks:
 - A bias towards 'moral emotions', and negative sentiment
 - A bias towards content about political out-groups
 - A bias towards false information.
- If the Recommendation System effects these biases, the instabilities could lead users in harmful directions.



Do Recommendation Systems lead users towards harmful content?

- Prima facie concern comes from theoretical models and simulations.
- But obviously, it must be tested on real social networks.
- Most studies are conducted externally to social media companies.
- But external methods all have limitations.
 - Population studies compare demographic groups with different Internet behaviours
 → confounding variables
 - User behaviour studies get data from volunteer social media users
 - *→* sampling problems
 - Robot user studies examine the consequences of following recommended links
 - → robots aren't real users
- The biggest problem: to test if a Recommendation Systems has causal effects on users, we
 must intervene on the Recommendation Systems—and that can only be done inside
 companies.



How companies study the effects of Recommendation Systems on users

- Social media companies are constantly trying out different versions of their Rec Sys on users, and picking the ones which are 'best', by their criteria.
- They use many criteria, but centrally they are looking for Rec Sys that maximise user engagement with their platform.
- Company-internal methods avoid the problems of external methods:
 - No confounding variables.
 - No sampling problems.
 - Studies are of real users, on real social media platforms.
 - Studies test proper causal hypotheses about Rec Sys effects.



A proposed 'fact-finding study'

- We propose a method for a government to work with a company, to ask whether its Rec
 Sys are moving users towards 'harmful content'.
 - We aim to trial this method in New Zealand, as a case study.
 - We are focussing on 'Terrorist and Violent Extremist Content' (TVEC), to fit in with this year's Christchurch Call workstream.
- Our fact-finding study augments **company's existing studies** of Rec Sys effects on users, with new metrics, that **measure users' engagement with 'harmful material'**.
 - Our focus is on metrics that gauge users' relationship towards TVEC.
 - The study is to be co-designed by the company and a group of independent experts.



The proposed 'fact-finding study' will ask two questions

1) Do different Recommendation Systems have different effects on users' relation towards harmful content? 2) Do Recommendation Systems that 'maximise user engagement' also drive users towards harmful content? Frances Haugen's recent revelations suggest the answer for Facebook's RecSys may be 'yes' in both cases.

But we can't rely on one-off disclosures based on unseen documents! We need a way of surfacing scientific findings about Rec Sys effects.

Publishing the results of the fact-finding study

Our proposal is that the results of a fact-finding study requested by a government are published in a scientific paper.

Our method is **safe**: doesn't compromise **company IP**

- delivers transparency about the effects of Recommendation Systems, not their internal workings
 doesn't compromise user privacy
 - measures of user behaviour are aggregated over large user groups

Twitter has recently published a paper describing exactly the kind of fact-finding study we envisage

 Huszár et al. "Algorithmic Amplification of Politics on Twitter", posted 21 October 2021



The community consultation project:

Community Consultative Processes for Grounded, Situated Harm-focused Responses

Question 1

How do the NZ communities who experience the most harm online from hateful expression, dangerous speech, and misinformation define those harms and their lived experiences of them?

Question 2

How can mediation, moderation, regulation, and categorisation as codeveloped and co-utilised tools mitigate against those harms and improve communities' experiences of online spaces?



Key method: meetings (hui) with community groups

Current online harm/disinformation in Aotearoa remains at high threat level

Researchers will be using word-of-mouth, trust-based channels for inclusion of participants in hui, to keep people safe.

Community meeting (hui)

Initial hui in June 2021.
Online community hui (zui) in Sept Oct. 2021. Preparation for larger hui's in 2022, focus on classification and categorisation for Aotearoa.



The law/policy project

investigates the legal/policy basis for the proposed fact-finding exercise.

For details, please see the written report!







THANK YOU

Contact: info@ceimia.org