



# AI and the Climate Crisis

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*Climate Informatics Paris, France* 

Keynote, October 4th 2019

### The Neural Net & Deep Learning Approach to AI: Amazing Progress but far from Human Level

- Brain-inspired
- Synergy of a large number of simple adaptive computational units
- Focus on distributed representations
  - E.g. word representations (Bengio et al NIPS'2000)
- View intelligence as arising of combining
  - an objective or reward function
  - an approximate optimizer (learning rule)
  - an initial architecture / parametrization
- End-to-end learning (all the pieces of the puzzle adapt to help each other)



# What did I have to do with Deep Learning?

- End-to-end training (1991-1998), understanding recurrent nets (TNN 1994), theory of distributed representations (NeurIPS 2000 to 2014)
- New methods to enable training of deeper networks + theoretical understanding
  - (Bengio et al NIPS'2006): pre-training stacks of auto-encoders before supervised training, greedy supervised and unsupervised pre-training
  - (Glorot & Bengio AISTATS'2010): initialization with near 1 e-values Jacobians
  - (Glorot & Bengio AISTATS'2011): importance of ReLU for training deep nets
- Beyond pattern recognition:
  - Progress in deep unsupervised models, generative models
    - (Vincent et al & Bengio 2008)++: denoising auto-encoders, self-supervised objectives
    - (Goodfellow et al & Bengio 2014): GANs = generative adversarial networks
  - Attention mechanisms, for arbitrary data structures, neural machine translation (Bahdanau et al & Bengio 2014)
  - Meta-learning (Bengio & Bengio 1991; many more in last 2 years)

# AI Scientists have a Responsibility

- How does my job fit with my values? Ask
  - How is this technology going to be used?
  - Who will benefit or suffer from it?
  - How much and what impact?
- I do not wish to work on
  - military applications, especially lethal autonomous weapons
  - ML to beat the stock market
  - increasing the efficiency of advertising or surveillance
  - increasing the efficiency of oil & gas industry, meat industry...

# AI for Social Good

- Beyond developing the next gadget
- Al is powerful, can be misused or bring much good
- Actionable items:
  - Favor ML applications which help the poorest countries, may help with fighting climate change, improve healthcare, education, etc.
  - AI Commons: an organization in construction, which will coordinate, prioritize and channel funding for such applications







HEC Renter for Human-Compatible Artificial Intelligence





Standards

Amir Banifatemi





## **Humanity's Biggest Crisis**



#### 2051-2080 Projected Change in Mean Temperature: December

Under the RCP8.5 scenario, relative to a baseline of 1976-2005





# **Climate Change in Quebec**

Climatic patterns (**temperature** and **flooding**) will change if emissions continue to rise.

#### 2051-2080 Projected Change in Total Precipitation: February

Under the RCP8.5 scenario, relative to a baseline of 1976-2005





Source: http://prairieclimatecentre.ca/2017/10/n ew-map-series-highlights-changescoming-to-canadas-climate/

# **Political Will**

- It may still technically be possible to limit warming to 1.5C if drastic action is taken now
- Politicians may act if there is sufficient priority given to climate change in the general population
- How do we convince those who are not yet and increase the priority they give to this issue?
  - $\circ$  there is a large intention-action gap





# Why is it Difficult to Act?

- 1. Fear of negative economic impacts of mitigation
  - Better estimate economic impacts, understand the high uncertainty
  - Yet adaptation costs > mitigation costs
  - Nash equilibrium between countries
- 2. Insufficient priority for the population and decision-makers
  - **Psychological factors** due to cognitive bias
- Lobby of a few large and powerful companies whose market will disappear, trying to delay the changes

# Mil

Enhance democratic tools!

#### Table A1

Estimation of Long-Term GDP Loss per Year Under Global Warming Scenarios (US\$ Billion/Year) to the Year 2100

4°C         3°C         2°C           World total         -23,149.18         -9,593.71         -5,659.47           United Kingdom         17.78         4.06         0.35           Germany         23.85         5.38         2.46           France         26.92         7.11         1.80           Italy         32.42         12.20         7.26           Canada         45.29         11.40         5.20           Rest of EU25         64.19         18.47         9.68				
United Kingdom         17.78         4.06         0.35           Germany         23.85         5.38         2.46           France         26.92         7.11         1.80           Italy         32.42         12.20         7.26           Canada         45.29         11.40         5.20		4° C	3°C	2° C
Germany23.855.382.46France26.927.111.80Italy32.4212.207.26Canada45.2911.405.20	World total	-23,149.18	-9,593.71	-5,659.47
France         26.92         7.11         1.80           Italy         32.42         12.20         7.26           Canada         45.29         11.40         5.20	United Kingdom	17.78	4.06	0.35
Italy         32.42         12.20         7.26           Canada         45.29         11.40         5.20	Germany	23.85	5.38	2.46
Canada 45.29 11.40 5.20	France	26.92	7.11	1.80
	Italy	32.42	12.20	7.26
Rest of EU25 64.19 18.47 9.68	Canada	45.29	11.40	5.20
	Rest of EU25	64.19	18.47	9.68



# **Psychological Factors**

- Out of sight, out of mind
- Far away in space (the catastrophes are always somewhere else)
- Far away in time
- Concepts are abstract
- Humans suffer from **cognitive biases** that are well-known: we underestimate the stakes!
- It is hard to compete with short-term economic issues
- Information bubbles/biases







# Climate Change + AI (CCAI) www.climatechange.ai

### Tackling Climate Change with Machine Learning

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# A menu for action:

### www.climatechange.ai





# A menu for action:

### www.climatechange.ai



topic you're interested in / working on

# A menu for action:

### www.climatechange.ai





### www.climatechange.ai

### Resources

### (readings, datasets, tools, and more)

Climate Change Al About V The Paper Resources Workshops V Newsletter V

#### Resources

This page presents some readings, datasets, and tools for the areas outlined in our paper, "Tackling Climate Change with Machine Learning." These resources include general references about climate change, as well as resources organized by section of the paper.

If you find that we are missing some important resources, please submit them through this form!

+	0. General
+	1. Electricity Systems
+	2. Transportation
+	3. Buildings and Cities
+	4. Industry
+	5. Farms and Forests
+	6. CO2 Removal
+	7. Climate Prediction
+	8. Societal Impacts
+	9. Solar Geoengineering
+	10. Tools for Individuals
+	11. Tools for Society
+	12. Education
+	13. Finance



Contact us at: climatechangeai@gmail.com

### www.climatechange.ai

### Resources (readings, datasets, tools, and more)

#### General

e.g. :

- Public Utility Data Liberation (PUDL) Project
- US independent system operators/regional transmission organizations (ISOs/RTOs): CAISO, ERCOT, ISO-NE, MISO, NYISO, PJM, and SPP

#### Variable Generation and Demand Forecasting

- Global Energy Forecasting Competition
- SubseasonalRodeo
- American Meteorological Society 2013-2014 Solar Energy Prediction Contest

#### Accelerated Science for Materials

- The Materials Project
- Inorganic Crystal Structure Database
- SciFinder (paid)
- UCI Machine Learning Repository datasets, e.g. "Concrete Compressive Strength"

#### Reducing Life-cycle Fossil Fuel Emissions

- Global methane data
- Also see satellite imagery listings under low-data settings



### **Tools for Individuals**

- Individual and household consumption optimization
  - ML can provide predictive tools and decision support
- Bias and individual behaviour change
  - e.g. using NLP for identifying bias in text, or helping users assess their own biases using chatbots

### **Tools for Society**

- Modeling collective behaviour
  - Game theory, mechanism design, multi-agent RL
- Managing tradeoffs
  - Operations research, multi-criteria optimization
- Markets and pricing (e.g. dynamic pricing of renewables)



### Education

- Analysis of learning
  - What do people understand about information when it is presented to them?
- Intelligent tutoring systems

### Finance

- Carbon markets and carbon pricing
- Identifying climate-related investment opportunities
- Auditing sustainability / carbon neutrality



- Optimizing energy resources
  - **Predicting output** from variable sources (solar, wind, hydro)
  - Estimating generated GHG to tax them and incite individuals and markets to change
  - Predict electricity demand to
    - optimize distribution to respond to fluctuating demand
    - optimize price to reduce consumption
  - Minimize losses and costs of transportation





- **Transportation**: predict demand (roads & public transport), improve urban infrastructure
- Construction: reduce electrical consumption of buildings
- **Industry:** optimize supply chains, maritime transport routes, waiting times







### Farms and forests

- Farms:
  - Real-time maps of Green House Gases emissions
  - Precision agriculture
- Forests:
  - Tracking deforestation
  - Automating afforestation

### Buildings

- Smart buildings:
  - tracking occupation, heating, lights
- Urban planning:
- Mila o to optimize travel routes





### CO2 Removal

- Direct air capture
  - ML can help the materials discovery process
- Sequestration
  - identify and characterize potential locations
  - monitor and maintain active sites

### Solar Geoengineering

- ML can help by:
  - modeling aerosols
  - defining objectives and targets





### **Climate modeling & prediction**

- "Downscaling" (linking models at different scales)
  - e.g. between a physics-based climate ensemble model and daily weather reports
- Identifying extreme weather events etc.
  - Very similar to object detection in images/video

### **Societal impacts**

Mila

- Disaster prediction and response
- Adaptation and mitigation
  - e.g. modeling effects of climate change on cities/infrastructure
- Computational sustainability
  - optimizing resource use to support life long-term





## HighRes-net MFSR by Recursive Fusion

### ELEMENT<sup>AI</sup>





Freddie Kalaitzis

Julien Cornebise

Zhichao Lin

Samira E. Kahou

Vincent Mich

i Yoshua Ben

ESTEC – Noordwijk | Kelvins Day | September 11, 2019



Low-resolution

Single Image Super-Resolution

High-res

### Low-res public satellites

- Sentinel 2 10 m
- Landsat 30 15 m
- PROBA-V 300 100 m
- \$ cheap / free
- → high revisit frequency







Low-res 1 Low-res 2 ... Low-res 9



High-res



Low-res 1 Low-res 2 ... Low-res 9

Super-res

High-res



Low-res 1 Low-res 2 ... Low-res 9

Super-res

High-res


















E





EAI

#### End-to-end training



#### ESA Super-Resolution challenge

#### Results

#### Leaderboard

Name	Final Score	
SuperPip	0.9474466476281652	
rarefin	0.9477450367529225	
ledzeppelin	0.9576339586408439	
luis	0.9626599626701344	
PIC4SeR	0.9721971077481462	
EscVM	0.9785367146826757	
SuperDeep	0.9807595119230784	

Name	Submissions	Best Score
rarefin	46	0.947388637793901
SuperPip	43	0.948844910272775
ledzeppelin	86	0.955016888143629
luis	9	0.964599100320855
PIC4SeR	15	0.971175807378147
EscVM	20	0.976616253152054
SuperDeep	38	0.981791472111074

## Synthesizing New Materials for Batteries using Deep Generative Models





Mila projects combining physics, chemistry and AI



Mila points of contact: Yoshua Bengio and Jordan Hoffmann

## **Societal Impacts of Climate Change**

#### ML has a great potential to support societal adaptation

- 1. Identifying and prioritizing the areas of highest risk, by using evidence of risk from historical data.
- 2. Providing annotation: Extracting actionable information or labels from unstructured raw data.
- 3. Promoting exchange: Making it easier to share resources and information to pool and reduce risk.







DV 05195

BY 58101



Mila point of contact: Kris Sankaran

# Using NLP to detect misinformation in media coverage of climate change

Term	Left-wing Articles	<b>Right-wing Articles</b>
global_ warming	human_activity (0.76)	climate_change (0.91)
	climate_crisis (0.75)	catastrophic_
	ACD (0.69)	unprecedented (0.80)
	man_made (0.68)	claim (0.75)
	scientific_	man_made (0.74)
	community (0.67)	warming_AGW (0.72)
	climate_	sublime (0.69)
	scientists (0.67)	Ehrlichs (0.68)
	scientific_	catastrophic (0.68)
	consensus (0.66)	groundless (0.66)
	human_activities (0.66)	scientists_agree (0.66)



Mila points of contact: Emi Baylor, Rosie Zhao (McGill), Sasha Luccioni

#### **Predicting Extreme Weather Events**



Mila points of contact: Evan Racah, Christopher Beckham, Tegan Maharaj, Samira Ebrahimi Kahou, Chris Pal



#### **Cloud Distribution Prediction using GANs**



Location, temperature, winds, humidity, reflectance



Image

Mila points of contact: Victor Schmidt, Mustafa Alghali, Kris Sankaran

#### **Measuring the Climate Emissions of ML**



+ Alex Lacoste (Element Al)

Mila point of contact: Victor Schmidt, Sasha Luccioni https://mlco2.github.io/impact

## Personalized Visualization of the Impact of Climate Change



- Allow people to educate themselves about the effects of climate change
- Do it in a **personalized** and **visual** way, to minimize cognitive biases
- Be positive: show that we can make the changes in time, by visualizing the impact of today's choices, such as the price of carbon
- Be **rigorous** and **scientific**, based on the most accurate climate models
- First prototype: a **website** that visualizes your home in 50 or 100 years, with the effect of floods, whose probabilities are calculated based on climate models

Mila point of contact: VICC team (Yoshua, Sasha, Karthik, Vahe, Victor, Gautier, Melisande, Ata, Yimeng…)



#### Personalized Visualization of the Impact of Climate Change





Mila point of contact: VICC team (Yoshua, Sasha, Karthik, Vahe, Victor, Gautier, Melisande...)



#### **Prospective Projects**

- Recognition of objects, rivers, plant cover, good conditions for planting trees in satellite imagery and aerial photogras, tree-planting drones
- Applying ML to mechanism/incentive design for policy research

• Using NLP to analyze climate disclosures

• Transfer learning for recognizing extreme weather events in weather data



## **Takeaway Messages**

- Machine learning, like all technology, does not always make the world a better place **but it can.**
- How you can get involved:
  - Learn: Identify how your skills may be useful (e.g. "*Tackling Climate Change with Machine Learning*" paper)
  - **Collaborate**: Find collaborators, who may be researchers, entrepreneurs, established companies, or policy-makers.
  - **Listen:** Listen to what your collaborators say is needed, and gather input more broadly as well to make sure your work will have the desired impact.
  - **Deploy**: Ensure that your work is deployed where its impact can be realized.



#### Half a million citizens at the Climat March in Montreal last Friday Sept 27th, 2019.



