



**Welcome to this session:**  
**Skills Bootcamp - Data Science for**  
**Social Good (Theory and Case**  
**Studies)**

**The session will start shortly...**

Questions? Drop them in the chat.  
We'll have dedicated moderators  
answering questions.



# Skills Bootcamp Data Science Housekeeping

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- The use of disrespectful language is prohibited in the questions, this is a supportive, learning environment for all - please engage accordingly. **(Fundamental British Values: Mutual Respect and Tolerance)**
- No question is daft or silly - **ask them!**
- There are **Q&A sessions** midway and at the end of the session, should you wish to ask any follow-up questions. We will be answering questions as the session progresses as well.
- If you have any questions outside of this lecture, or that are not answered during this lecture, please do submit these for upcoming Academic Sessions. You can submit these questions here: **Questions**

# Skills Bootcamp Data Science Housekeeping

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- For all **non-academic questions**, please submit a query: [www.hyperiondev.com/support](http://www.hyperiondev.com/support)
- Report a safeguarding incident: [www.hyperiondev.com/safeguardreporting](http://www.hyperiondev.com/safeguardreporting)
- We would love your feedback on lectures: [Feedback on Lectures.](#)
- Find all the lecture **content** in your [Lecture Backpack](#) on GitHub.
- If you are hearing impaired, kindly use your computer's function through Google chrome to enable captions.

# Safeguarding & Welfare

We are committed to all our students and staff feeling safe and happy; we want to make sure there is always someone you can turn to if you are worried about anything.

If you are feeling upset or unsafe, are worried about a friend, student or family member, or you feel like something isn't right, speak to our safeguarding team:



Ian Wyles  
Designated Safeguarding  
Lead



Simone Botes



Nurhaan Snyman



Rafiq Manan



Ronald Munodawafa



Tevin Pitts

Scan to report a  
safeguarding concern



or email the Designated  
Safeguarding Lead:  
Ian Wyles

[safeguarding@hyperiondev.com](mailto:safeguarding@hyperiondev.com)

# Skills Bootcamp Progression Overview

## ✓ Criterion 1 - Initial Requirements

Specific achievements **within the first two weeks** of the program.

To meet this criterion, students need to, by no later than **01 December 2024 (C11)** or **22 December 2024 (C12)**:

- **Guided Learning Hours (GLH):** Attend a **minimum of 7-8 GLH per week** (lectures, workshops, or mentor calls) for a total minimum of **15 GLH**.
- **Task Completion:** Successfully complete the **first 4 of the assigned tasks**.

## ✓ Criterion 2 - Mid-Course Progress

Progress through the successful completion of tasks **within the first half** of the program.

To meet this criterion, students should, by no later than **12 January 2025 (C11)** or **02 February 2025 (C12)**:

- **Guided Learning Hours (GLH):** Complete at least **60 GLH**.
- **Task Completion :** Successfully complete the **first 13 of the assigned tasks**.

# Skills Bootcamp Progression Overview

## ✓ Criterion 3 – End-Course Progress

Showcasing students' progress nearing the completion of the course.

To meet this criterion, students should:

- **Guided Learning Hours (GLH):** Complete the **total minimum required GLH**, by the **support end date**.
- **Task Completion : Complete all mandatory tasks**, including any necessary resubmissions, by the end of the bootcamp, **09 March 2025 (C11)** or **30 March 2025 (C12)**.

## ✓ Criterion 4 - Employability

Demonstrating progress to find employment.

To meet this criterion, students should:

- **Record an Interview Invite:** Students are required to record proof of invitation to an interview by **30 March 2025 (C11)** or **04 May 2025 (C12)**.
  - **South Holland Students** are required to proof and interview by **17 March 2025**.
- **Record a Final Job Outcome :** Within 12 weeks post-graduation, students are required to record a job outcome.



## Learning Outcomes

- ❖ **Define and explain the role of data science in social good initiatives**, including areas such as public health, education, poverty reduction, and humanitarian aid.
- ❖ **Explain theoretical frameworks** for **designing data-driven solutions** that prioritize equity, sustainability, and ethical responsibility.
- ❖ **Analyse case studies** where **data science has driven positive social impact**, understanding methodologies, data sources, and ethical considerations.
- ❖ **Identify and evaluate ethical concerns** in social good applications, including issues of bias, privacy, and unintended consequences.
- ❖ **Assess potential challenges and propose solutions** to ensure responsible and impactful use of data science for social good.

# Lecture Overview

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- Data Science for Social Good
- Theoretical Frameworks
- Case Studies and Real World Applications
- Ethical Concerns & Solutions







# Which of the following best defines “Data Science for Social Good”?

- A. Using data to maximize profit in businesses
- B. Applying data science techniques to address societal challenges
- C. Automating social services using AI
- D. Collecting public data without ethical concerns



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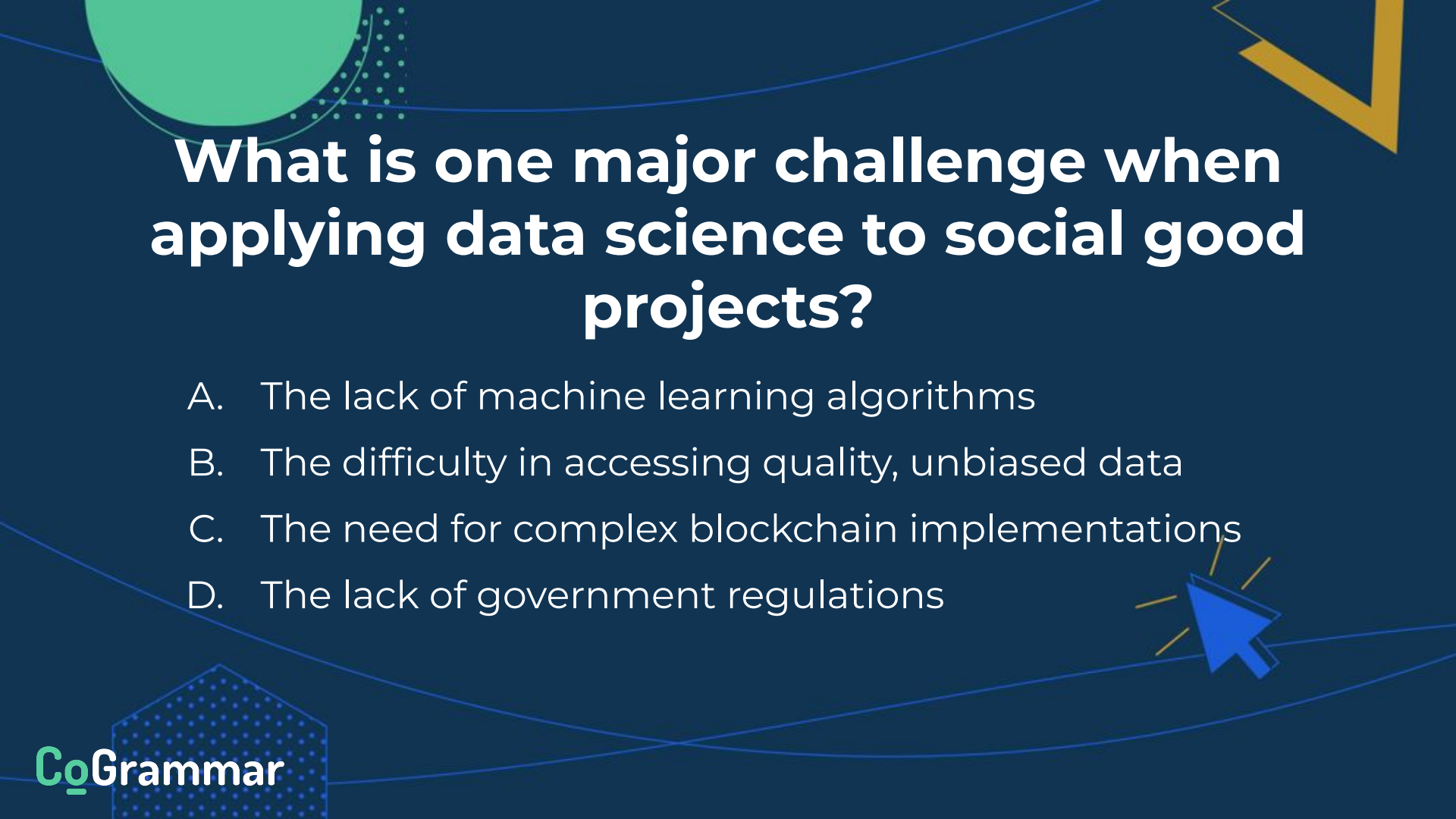
# Which area is not commonly associated with data science for social good?

- A. Public health
- B. Disaster response
- C. Predicting stock prices
- D. Education accessibility



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# What is one major challenge when applying data science to social good projects?

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- B. The difficulty in accessing quality, unbiased data
- C. The need for complex blockchain implementations
- D. The lack of government regulations



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# Data Science for Social Good

Data science has transformed industries like finance, healthcare, and technology. But can it also be a force for social good? Governments, NGOs, and research institutions increasingly use data-driven approaches to tackle challenges like poverty, climate change, and public health crises. However, implementing these solutions effectively requires overcoming issues like bias, accessibility, and ethical concerns.

# Data Science for Social Good

- Have you seen examples where **data science** has been used to **solve social problems**?
- What **challenges** do you think arise when applying data science in real-world social contexts?

# Data Science for Social Good

The application of data science techniques and methodologies to tackle social issues and improve the well-being of individuals and communities

- ❖ It aims to move beyond simply extracting insights from data and focuses on **creating solutions** that can **improve people's lives** and **address systemic problems**.
- ❖ **Nonprofit organizations, civic agencies, and public sectors** recognize data science's immense potential to **address social challenges** and **drive positive societal change**.

Have a look at [GoogleAI's Social Impact page](#) to see how they've been using AI for Social Good.

# Data Science for Social Good

## ❖ Key Areas of Impact:

- **Healthcare:** Predicting disease outbreaks, optimising hospital resources.

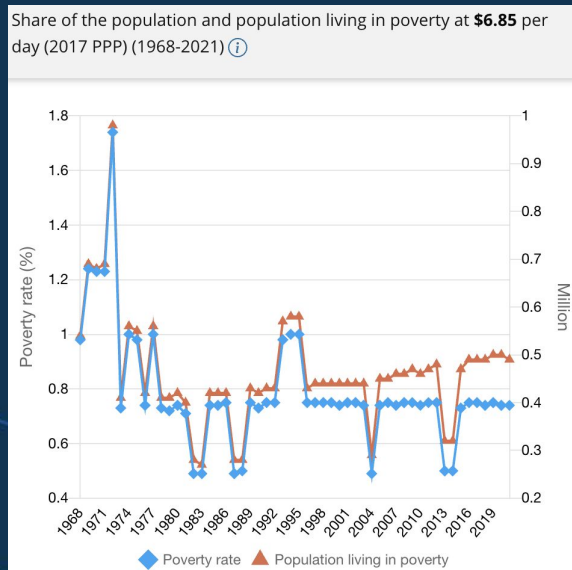


Source: [WHO's AI for Health Approach](#)

# Data Science for Social Good

## ❖ Key Areas of Impact:

- **Poverty & Inequality:** Identifying underserved communities, targeting aid effectively.

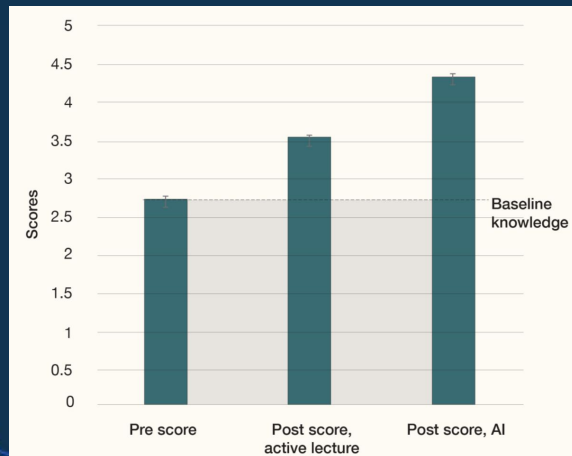


The [World Bank](#) has been using various tools to collect and share useful data with the aim to help track and understand poverty, its causes and its impacts.

# Data Science for Social Good

## ❖ Key Areas of Impact:

- **Education:** Improving learning outcomes through personalised AI tutors.



A comparison of mean post-test performance between students taught with the active lecture and students taught with the AI tutor. Dotted line represents students' mean baseline knowledge before the lesson (i.e. the pre-test scores of both groups).  
Source: "AI Tutoring Outperforms Active Learning," Gregory Kestin, Kelly Miller, Anna Klales, Timothy Milbourne, Gregorio Ponti

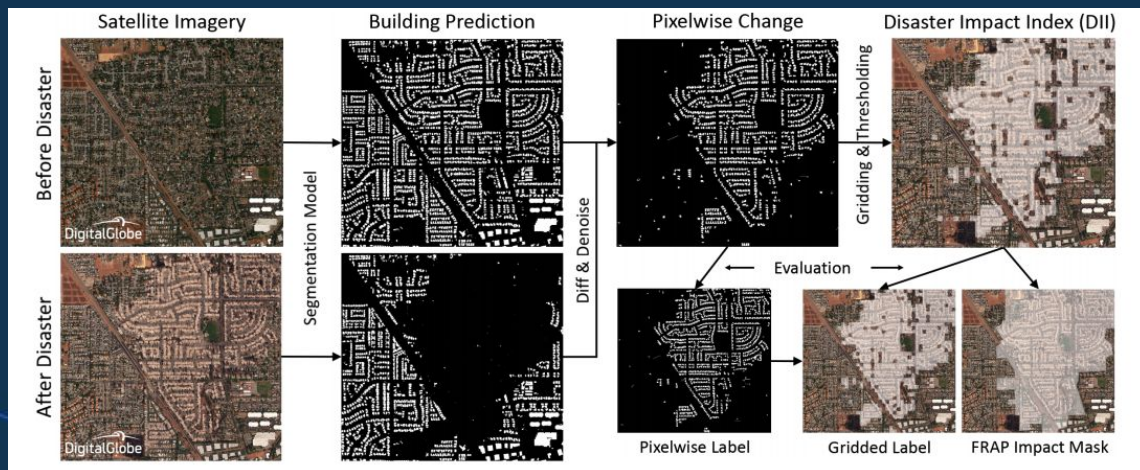
A [Harvard study](#) found that engagement doubled with students who were assisted by an AI tutor.



# Data Science for Social Good

## ❖ Key Areas of Impact:

- **Disaster Response:** Using satellite imagery & AI to detect and mitigate natural disasters.

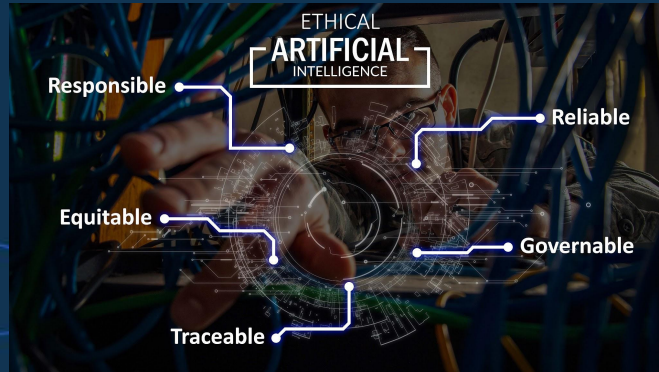


Source: [Detecting Disaster Damage with AI](#)

# Theoretical

## ❖ How to Design Data Solutions for Social Good:

- **Fairness & Bias:** Avoiding bias in algorithms to ensure equitable outcomes.
- **Transparency & Interpretability:** Making AI-driven decisions understandable.
- **Sustainability:** Long-term benefits rather than short-term fixes.



# Theoretical

- ❖ We can aim to ensure that our tools and methodologies are aligned with the overarching goals of Data Science for Social Good using these frameworks as guidelines:
  - [Data4Good Framework](#) – balancing impact with ethics by laying out a roadmap for responsible data sharing in the UAE.
  - [AI Ethics Guidelines](#) – proposed by organizations like the EU, UNESCO to provide a global resource for policymakers, regulators, academics, the private sector and civil society to find solutions to the most pressing challenges posed by Artificial Intelligence.

# Let's Breathe!

Let's take a small break  
before moving on to  
the next topic.



# Case Studies

## ❖ Predicting Crime in Chicago Using AI

Researchers at the University of Chicago developed an algorithm to predict crime by analysing the time and spatial coordinates of discrete events. The model successfully predicted crimes across various U.S. cities, including Chicago, Atlanta, and Los Angeles. However, it also **revealed biases in police responses**, indicating **increased attention in wealthier neighborhoods** at the expense of less advantaged areas.



# Case Studies

## ❖ NASA's AI for Disaster Response

NASA, in collaboration with IBM Research, developed an **open-source artificial intelligence model** to support a variety of weather and climate applications. This AI model leverages NASA's extensive satellite data to improve the resolution of regional and local weather and climate models. It supports applications such as **tracking changes in land use, monitoring disasters, and predicting crop yields** worldwide.



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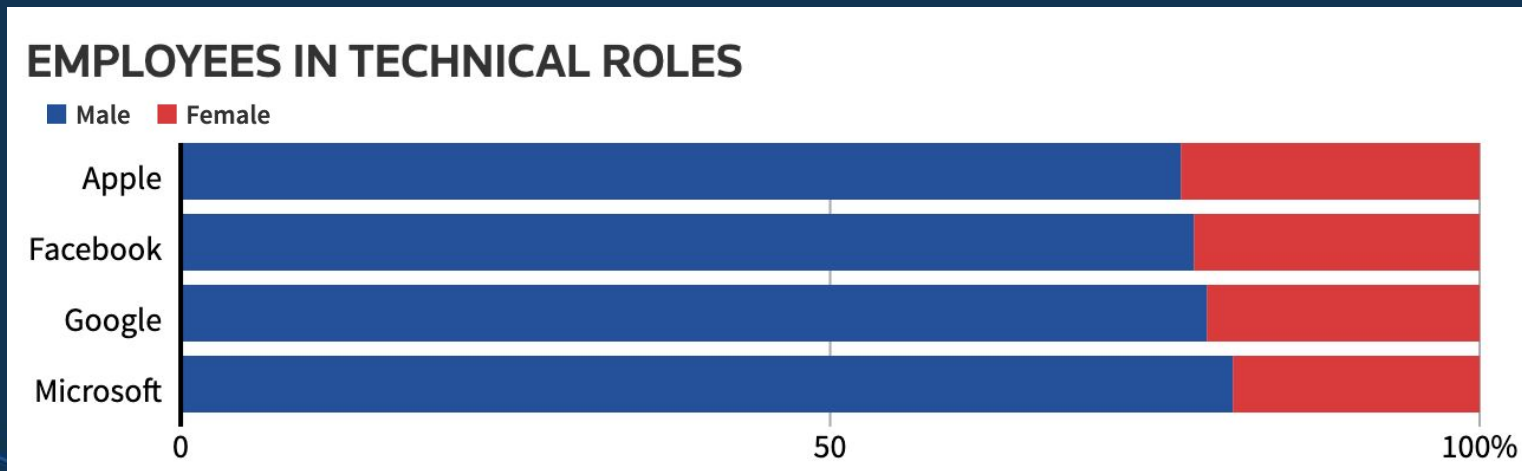
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# Ethical Considerations

- ❖ **Bias in AI Models**: How biased datasets can reinforce discrimination.



[Amazon had to scrap its recruiting tool](#) after observing that the model was biased against women since it had been trained on biased hiring data.

# Ethical Considerations

- ❖ **Privacy & Consent**: Handling personal data responsibly.



Governments, corporations, universities and other researchers frequently rely on data collected from marginalized populations, yet these communities rarely benefit equitably from the insights or policies derived from their information. This imbalance raises significant ethical concerns about privacy, exploitation, and agency. [This guideline](#) was released to aid in responsible and equitable data sharing.

# Ethical Considerations

- ❖ **Transparency & Accountability:** Ensuring AI decisions can be explained.

## Requirements of AI transparency



Explainability



Interpretability



Accountability

Whenever AI is being used, it's important that decisions made by the system are explainable (we know **why** the decision was made), interpretable (we know understand **how** the system works) and accountable (we can **correct mistakes** and the system **learns from mistakes**). If not, systems could be biased, unreliable or untrustworthy.



# Future of Data Science

## ❖ Emerging Technologies in AI & Data Science:

- Federated Learning for privacy-preserving AI.
- Explainable AI (XAI) for better transparency.
- AI-driven climate solutions (carbon footprint tracking).

## ❖ How Students Can Get Involved

- Open-source projects for social good (DataKind, UNICEF's AI initiatives).
- Collaborate with NGOs, research groups working on AI for Good.






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- A. The presence of bias in datasets
- B. The high cost of implementing AI solutions
- C. The lack of advanced AI models
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- B. Prioritizing speed over accuracy
- C. Using explainable AI models to detect bias
- D. Removing ethics considerations from models



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# What is one way data science has improved disaster response?

- A. Creating better insurance policies for affected individuals
- B. Reducing government involvement in emergency situations
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## Summary

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- ★ Data science has the power to **drive social change** when used **responsibly**.
- ★ Case studies show its impact on **health, disaster response, education, and poverty reduction**.
- ★ **Challenges like bias, privacy, and scalability** must be carefully managed.
- ★ **Ethical considerations** should always be at the core of social good applications.



# CoGrammar

## Q & A SECTION

**Please use this time to ask  
any questions relating to the  
topic, should you have any.**

# Thank you for attending



**CoGrammar**



Department  
for Education