

# DATA & ETHICS: “WITH GREAT POWER COMES GREAT RESPONSIBILITY”



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# ETHICS, ETHICS EVERYWHERE

## THERE ARE WELL-KNOWN CASES OF ETHICAL DILEMMAS

- Cambridge Analytica Facebook data
- NYC "My City" chatbot
- Amazon hiring algorithm
- Clearview AI image database
- Apple AirTags
- Target predictive marketing model



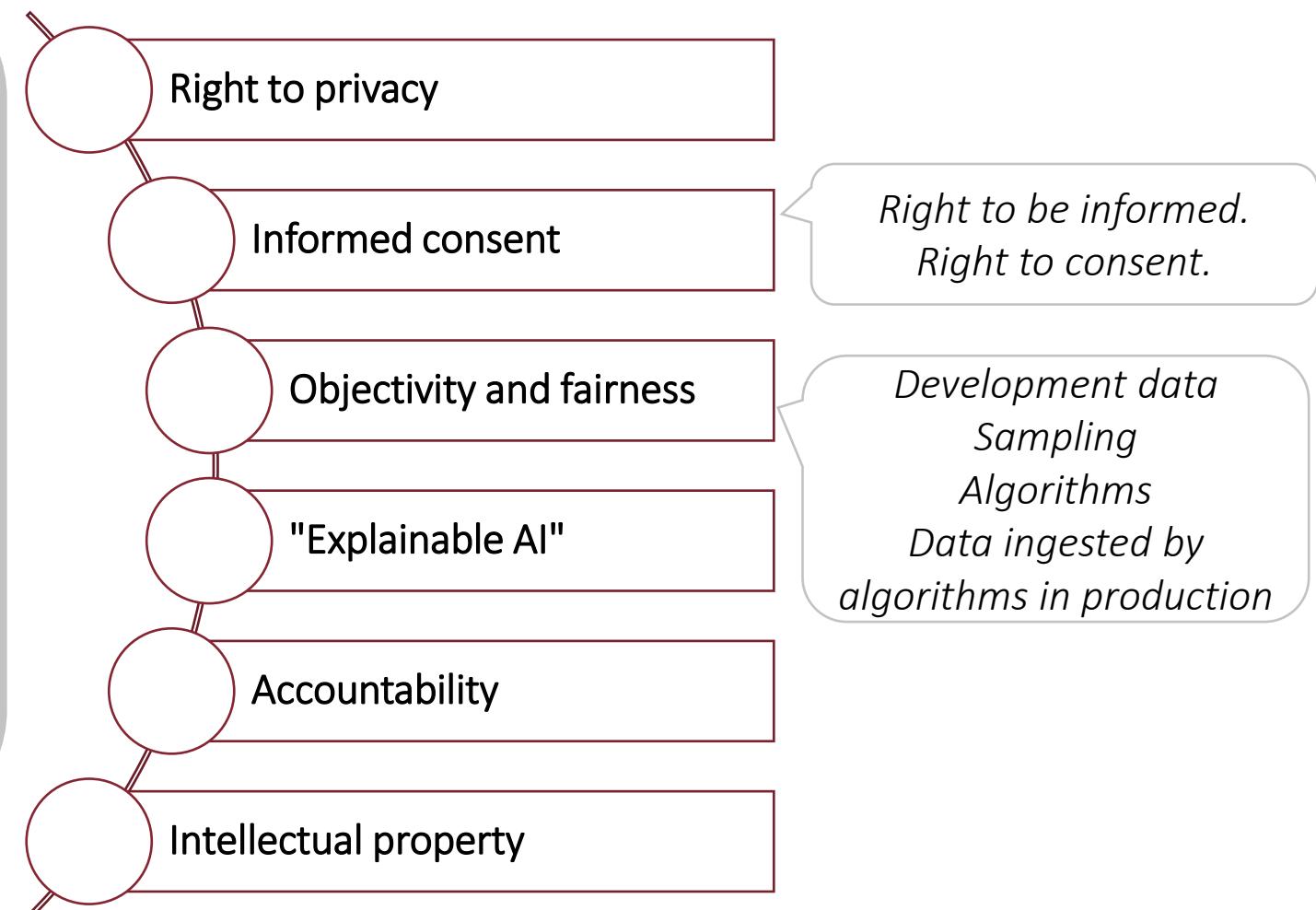
# ETHICS IN DATA AND ANALYTICS

## SOME HOT TOPICS

**Responsible AI:** The practice of designing, developing, and deploying artificial intelligence systems in ways that are ethical, transparent, fair, and aligned with human values and societal well-being.

**Responsible data:** The ethical and secure handling of data to ensure privacy, fairness, transparency, and accountability throughout its lifecycle.

\*According to Microsoft Copilot



# IMPLEMENTATIONS: ETHICS IN DATA AND ANALYTICS

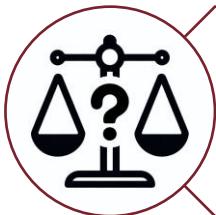
## CURRENT LANDSCAPE

- **Regulations:** Data, algorithms
- **Research ethics review**
- **Industry standards and frameworks**
- **Certifications**
- **Professional guidelines and codes of ethics**
- **Organizational policies and practices**
- **Working groups, conferences, etc.**

- *Respect for stakeholders: data subjects, clients, colleagues/collaborators, users, and the general public*
- *Professional competency*
- *Respect for the profession*
- *Obligation to assume responsibility as a professional*
- *Obligation to do no harm*
- *Honesty/transparency*
- *Fairness*
- *Privacy/confidentiality*

# THE NATURE OF ETHICAL DILEMMAS

## IF IT WERE EASY, EVERYONE WOULD DO IT



Ethical vs. Legal



Conflicting values



Not always “black and white”

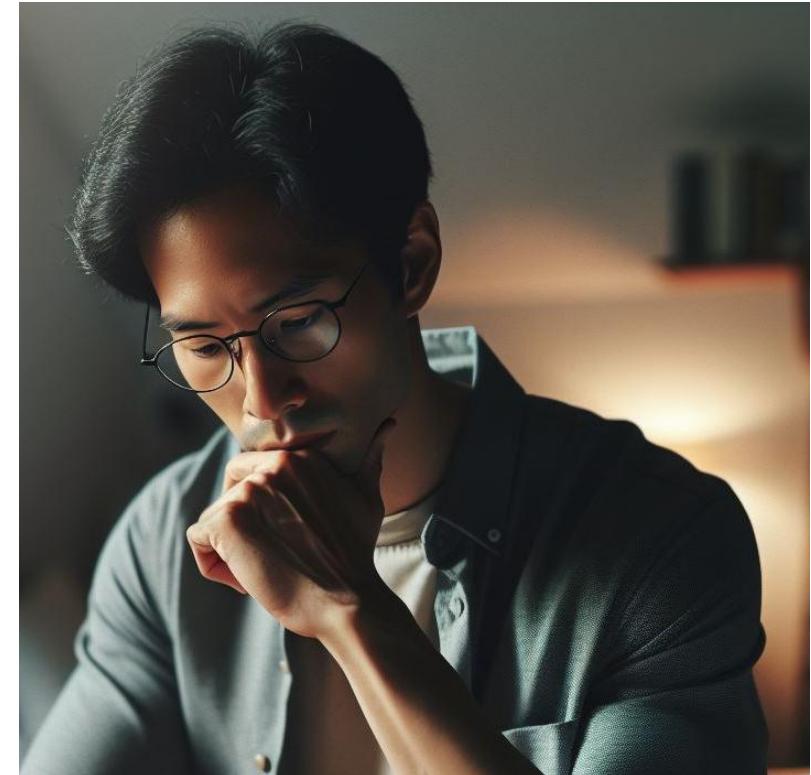


Require careful thought and critical thinking

# ETHICAL ISSUES FOR STATISTICIANS AND DATA SCIENTISTS

## IT'S ABOUT WHAT YOU DO, NOT WHAT HAPPENS TO YOU

- "Can" vs. "should"
  - Knowing when to do something
  - when NOT to do something.
- Different aspects: Objective, data, algorithms, use.
- Fairness vs. bias
  - Global/social bias vs. personal bias
  - Theory vs. reality
  - Open biases vs. hidden biases
  - Bias in: (1) data, (2) algorithms, (3) users
- Intentionality and justification
  - Do you know what you are doing?
  - How can you demonstrate it?
- Building trust
  - How do we demonstrate our commitment to ethical principles?



# ETHICAL ISSUES FOR STATISTICIANS AND DATA SCIENTISTS

## THERE IS A LOT TO CONSIDER IN EACH OF THE FOUR ASPECTS

### Objective of the solution

- Should we develop it?
- Targeting vs. fairness
- Experimentation vs. manipulation
- Details of the intended uses
- Potential unintended uses
- Scope of its impact

### Data: In development and in production

- Anonymization and encryption are necessary but not sufficient to protect privacy
- Use of data only for defined purposes
- Data minimization
- Inherent biases in modern data sources
- Intellectual property in data

### Algorithms

- The selection of the technique: Performance vs. transparency
- Model architecture: variables/features, segmentation, treatments, etc.
- Reducing the cost of misinformation
- Incorporating fairness and eliminating biases

### Use of the solution

- Unintended consequences of intended uses
  - Risks and costs of misinformation
  - Consequences of naive uses
- Uses beyond those originally intended
  - Naivety? Manipulation? Malice?
- Perpetuating or worsening bias
- Intellectual properties and ownership

# HOW DO I PUT ALL THIS INTO PRACTICE?

TAKE THE TIME AND EFFORT TO DO THE RIGHT THING BY RESPECTING GENERAL PRINCIPLES



Do no harm.

Be fair.

Be honest and  
transparent.

Take responsibility.

Respect and  
protect others.

- Be deliberate and intentional in everything you do, down to the smallest details.
- Proactively identify and disclose any conflicts of interest.
- Be aware of all shortcomings.
- Document all decisions and justifications.
- Proactively monitor and correct methodologies, practices, and products: biases, misinformation.
- Take responsibility for own actions and decisions.

# QUESTIONS TO USE AS A STARTING POINT TO GUIDE OUR DECISIONS AND ACTIONS FROM AN ETHICAL PERSPECTIVE



Is it consistent with laws, regulations, and standards?



What are the risks?



Whom does it compromise?



To whom does the information belong?



Whom does it put at risk?



What are the possible consequences?



Is it moral? For whom?



Whom does it benefit?



Does it work as intended for all stakeholders?

Legal/regulatory • Moral • Financial • Reputational • Physical • Psychological • Direct/indirect • Etc.

How do you know? How have you demonstrated it?



Michiko I. Wolcott, Principal Consultant  
Msight Analytics

[michiko.wolcott@msightanalytics.com](mailto:michiko.wolcott@msightanalytics.com)

 @Michiko\_Wolcott  
 @michiko\_wolcott

 [linkedin.com/in/michikowolcott/](https://www.linkedin.com/in/michikowolcott/)

 @msightanalytics  
 @msightanalytics  
 @michiko-wolcott

# REFERENCES

(EXAMPLES; NOT EXHAUSTIVE)

# EXAMPLES OF REGULATIONS AND STANDARDS

Jurisdiction	Name	URL
European Union	General Data Protection Regulation (GDPR)	<a href="https://gdpr.eu/tag/gdpr/">https://gdpr.eu/tag/gdpr/</a>
European Union	EU Artificial Intelligence Act	<a href="https://artificialintelligenceact.eu/">https://artificialintelligenceact.eu/</a>
United Kingdom	UK GDPR and Data Protection Act	<a href="https://ico.org.uk/for-organisations/data-protection-and-the-eu/data-protection-and-the-eu-in-detail/the-uk-gdpr/">https://ico.org.uk/for-organisations/data-protection-and-the-eu/data-protection-and-the-eu-in-detail/the-uk-gdpr/</a>
California, USA	California Consumer Privacy Act (CCPA)	<a href="https://oag.ca.gov/privacy/ccpa">https://oag.ca.gov/privacy/ccpa</a>
Mexico	Federal Law on the Protection of Personal Data Held by Private Parties (LFPDPPP)	<a href="https://www.gob.mx/idesol/documentos/ley-federal-de-proteccion-de-datos-personales-en-posesion-de-los-particulares">https://www.gob.mx/idesol/documentos/ley-federal-de-proteccion-de-datos-personales-en-posesion-de-los-particulares</a>
Canada	Personal Information Protection and Electronic Documents Act (PIPEDA)	<a href="https://www.priv.gc.ca/en/privacy-topics/privacy-laws-in-canada/the-personal-information-protection-and-electronic-documents-act-pipeda/">https://www.priv.gc.ca/en/privacy-topics/privacy-laws-in-canada/the-personal-information-protection-and-electronic-documents-act-pipeda/</a>
Brazil	General Data Protection Law (LGPD)	<a href="https://www.gov.br/esporte/pt-br/acao-a-informacao/lgpd">https://www.gov.br/esporte/pt-br/acao-a-informacao/lgpd</a>
Argentina	Personal Data Protection Law	<a href="https://servicios.infoleg.gob.ar/infolegInternet/anexos/60000-64999/64790/norma.htm">https://servicios.infoleg.gob.ar/infolegInternet/anexos/60000-64999/64790/norma.htm</a>
India	Digital Personal Data Protection Act (DPDP Act)	<a href="https://egazette.gov.in/WriteReadData/2023/248045.pdf">https://egazette.gov.in/WriteReadData/2023/248045.pdf</a>

# EXAMPLES OF FRAMEWORKS AND STANDARDS

Name	Entity	URL
International Compilation of Human Research Standards	Office of Human Research Protections, U.S. Health and Human Services (HHS)	<a href="https://www.hhs.gov/ohrp/international/compilation-human-research-standards/index.html">https://www.hhs.gov/ohrp/international/compilation-human-research-standards/index.html</a>
Belmont Report: Ethical Principles and Guidelines for the Protection of Human Subjects of Research	Office of Human Research Protections, U.S. Department of Health and Human Services (HHS)	<a href="https://www.hhs.gov/ohrp/regulations-and-policy/belmont-report/index.html">https://www.hhs.gov/ohrp/regulations-and-policy/belmont-report/index.html</a> <a href="https://www.hhs.gov/sites/default/files/informed-belmont-spanish.pdf">https://www.hhs.gov/sites/default/files/informed-belmont-spanish.pdf</a>
Privacy by Design	Information and Privacy Commissioner of Ontario	<a href="https://en.wikipedia.org/wiki/Privacy_by_design">https://en.wikipedia.org/wiki/Privacy_by_design</a>
Ethics Guidelines for Trustworthy AI	European Commission	<a href="https://digital-strategy.ec.europa.eu/en/library/ethics-guidelines-trustworthy-ai">https://digital-strategy.ec.europa.eu/en/library/ethics-guidelines-trustworthy-ai</a>
AI Risk Management Framework	National Institute of Standards and Technology (NIST)	<a href="https://www.nist.gov/itl/ai-risk-management-framework">https://www.nist.gov/itl/ai-risk-management-framework</a>
ISO/IEC 42001	International Standards Organization (ISO)	<a href="https://www.iso.org/standard/81230.html">https://www.iso.org/standard/81230.html</a>
IEEE 7000	IEEE Standards Association	<a href="https://standards.ieee.org/ieee/7000/6781/">https://standards.ieee.org/ieee/7000/6781/</a>
Artificial Intelligence Governance Professional (AIGP) Certification	International Association of Privacy Professionals (IAPP)	<a href="https://iapp.org/certify/aigp/">https://iapp.org/certify/aigp/</a>

# EXAMPLES OF PROFESSIONAL GUIDELINES AND CODES OF ETHICS

Organization	Name	URL
American Statistical Association (ASA)	Ethical Guidelines for Statistical Practice	<a href="https://www.amstat.org/your-career/ethical-guidelines-for-statistical-practice">https://www.amstat.org/your-career/ethical-guidelines-for-statistical-practice</a>
American Statistical Association (ASA)	Statement on Ethical AI Principles for Statistical Practitioners	<a href="https://www.amstat.org/docs/default-source/amstat-documents/asa-statement-on-ethical-ai-principles-for-statistical-practitioners.pdf">https://www.amstat.org/docs/default-source/amstat-documents/asa-statement-on-ethical-ai-principles-for-statistical-practitioners.pdf</a>
Association for Computing Machinery (ACM)	Code of Ethics	<a href="https://www.acm.org/code-of-ethics">https://www.acm.org/code-of-ethics</a>
Institute of Management Consultants USA (IMC USA)	Code of Ethics	<a href="https://imcusa.org/about/ethics/code-of-ethics/">https://imcusa.org/about/ethics/code-of-ethics/</a>

# EXAMPLES OF ETHICAL CASES

- “Facebook-Cambridge Analytica Data Scandal”: [https://en.wikipedia.org/wiki/Facebook%20Cambridge\\_Analytica\\_data\\_scandal](https://en.wikipedia.org/wiki/Facebook%20Cambridge_Analytica_data_scandal)
- “Clearview AI”: [https://en.wikipedia.org/wiki/Clearview\\_AI](https://en.wikipedia.org/wiki/Clearview_AI)
- “Judge denies Apple’s attempt to dismiss a class-action lawsuit over AirTag stalking”: <https://apnews.com/article/apple-airtags-stalking-lawsuits-e59166988920c4ba1e82956ea85c1677>
- “Insight - Amazon scraps secret AI recruiting tool that showed bias against women”: <https://www.reuters.com/article/world/insight-amazon-scaps-secret-ai-recruiting-tool-that-showed-bias-against-women-idUSKCN1MK0AG/>
- “NYC’s AI Chatbot Tells Businesses to Break the Law”: <https://themarkup.org/news/2024/03/29/nycs-ai-chatbot-tells-businesses-to-break-the-law>
- “How Companies Learn Your Secrets” (Target predictive modeling): <https://www.nytimes.com/2012/02/19/magazine/shopping-habits.html?pagewanted=1& r=1&hp#>
- “Artificial Intelligence Bot Becomes Racist And Homophobic After Learning From Humans”: <https://oecd.ai/en/incidents/16537>
- “New York lawyers sanctioned for using fake ChatGPT cases in legal brief”: <https://storage.courtlistener.com/recap/gov.uscourts.nysd.575368/gov.uscourts.nysd.575368.31.0.pdf>
- “Racial Bias Found in a Major Health Care Risk Algorithm”: <https://www.scientificamerican.com/article/racial-bias-found-in-a-major-health-care-risk-algorithm/>