**Introduction**

- The future of artificial intelligence is far from certain, but AI presents many immediate concerns and research opportunities for ethical decision making in AI (Bauer & Dubljević 2019).
- Using AI to reverse-engineer the brain may aid researchers in understanding both human and artificial intelligence, but there is little empirical ethics research into which areas are most relevant or feasible.
- As in other areas of neuroethics scholarship, there is a need to separate urgent from speculative issues (Racine et al. 2017).

**Methods**

- Building on prior work (Dubljević et al. 2021), we conducted nine semi-structured interviews with computer science and engineering experts specializing in AI near Raleigh, North Carolina.
- We seek to discern the issues, concerns, and directions for future research in AI around nine broad themes, from the R.L. Rabb Symposium (AI in Society 2021) and five from the National Academy of Engineering’s Grand Challenges (NAE). Two of these themes were of particular interest:
  - Integrating ethics into AI decision making
  - Reverse-engineering the brain
- We used two methods of data analysis in order to get a comprehensive grasp on what experts in the field felt were pertinent areas as AI advances:
  - Qualitative Interview Methodology (Timmermans & Tavori 2012)
    - Data analysis conducted concurrently with data collections
    - Codes were developed via abductive analysis, and inter-coder reliability was high (84.38%)
  - Delphi Methodology (Okoli & Pawlowski 2004)
    - First round (interviews)
      - Asked interviewees to list at least three topics for each theme
      - Demographic data was collected (gender, disciplinary background, and location)
    - Second round (survey)
      - Topics from the first round were consolidated into manageable, short sentences
      - Asked respondents to select three important topics and three that are not feasible
    - Third round (survey)
      - Topics that were selected by a majority of respondents were included
      - Asked respondents to rank each topic based on importance, desirability, and feasibility

**Qualitative Interview Quotes**

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<tr>
<th>Data vs. Algorithms</th>
<th>Reverse-engineering the brain</th>
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<tbody>
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**Delphi (First Round Topics)**

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**Conclusion**

- Given the open-ended potential for AI to aid both human and artificial intelligence, it is imperative to understand the best practices in AI research and the implications for neuroethics.
- Our study shows that there is a strong desire among AI researchers to keep ethics and public good in mind when training AI systems to reverse-engineer the brain.

**References**


**Figure 1**

- Ethical Perspectives from AI Experts on Reverse-Engineering the Brain

- Data vs. Algorithms
  - “Anybody who falls outside of those norms is going to have a higher chance of being misclassified. And falling outside of those norms is, of course, relative to the data set that it’s trained on” (Page 6)
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- Reverse-engineering the brain
  - “There’s a lot of excitement about artificial neural networks, but how closely these actually teach us anything about actual neurons is […] a real question that hasn’t been addressed very well” (Page 33)
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  - Our study shows that there is a strong desire among AI researchers to keep ethics and public good in mind when training AI systems to reverse-engineer the brain.

**Table 1**

- Integrating ethics into AI decision making
  - Biases built into algorithms and training data
  - Which ethical theories to introduce or use in AI
  - Equity and fairness in machine learning and data sets
  - Representational issues in the data inputted into AI
  - False positives (identified by the AI) that lead to negative outcomes

- Reverse-engineering the brain
  - Leaving the human out of the picture
  - Use AI as a neuroscience research aid (e.g. using AI for behavioral research or with EEGs to monitor sleep cycles)
  - Use neuroscience research to aid AI research
  - Brain-computer interfaces
  - Neurodiversity among human brains and/or embedding a hegemony of averages into a reverse-engineered brain