



Mapping the Public Debate on Ethical Concerns: Algorithms in Mainstream Media

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Mapping the Public Debate on Ethical Concerns: Algorithms in Mainstream Media

Algorithms are in the mainstream media news on an almost daily basis. Their context is invariably artificial intelligence and machine learning decision making. In media articles, algorithms are described as powerful, autonomous actors that have a capability of producing actions that have consequences. Despite a tendency to deification, the prevailing critique of algorithms focusses on ethical concerns raised by decisions resulting from algorithmic processing. However, this paper proposes that the ethical concerns discussed are limited in scope and it is not clear which concerns dominate the debate. The research presented in this paper contributes the first systematic mapping study of articles appearing in leading UK national papers from the perspective of widely accepted ethical concerns. The UK context is important because of UK public policy initiatives around artificial intelligence. In academic literature, the key ethical concerns have been well documented and numerous models have been developed. To review the media content from the perspective of ethical concerns, this paper uses the synthesised conceptual map of ethical concerns developed by Mittelstad et al. Given the widespread use of that framework, as evident through citations, this paper's contribution is also an important illustration and experiment using that conceptual map.

Keywords: Ethics, Algorithms, Machine Learning, Artificial Intelligence, Bias, Governance

1 Introduction

'Algorithms' are in the mainstream media (the so-called fourth estate (Newman et al., 2012)) news. Their description follows a predictable, almost algorithmically scripted drama of two acts that invariably conjures a deification of the algorithm (Bogost, 2015). In this seductive drama elucidated with much eloquence by Ziewitz (2016) we are first introduced to an algorithm as a powerful autonomous actor with a capability of producing actions that have consequence. Examples of consequences may be derived

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3 from medicine, social media, policing or even employment. In the last act of this drama
4 the autonomous, agency like characteristics of algorithms are enhanced further, through
5 descriptions of bias, opacity, mistakes and corporate misdemeanour. The drama
6 concludes by pointing out a need for governance, regulation and a new logic for ethics.
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8 In this manner, the public discourse in the mainstream media about algorithms unfolds.
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10 Recall one week in May 2018 in the UK, where the BBC reported on Amnesty
11 International exposing a Metropolitan Police database as racially discriminatory; a
12 facial recognition system used by South Wales police throwing up false positives and
13 the data profiling company Cambridge Analytica under scrutiny by the Information
14 Commissioner (Andrews, 2018).
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26 One newspaper article demonstrates this drama in an almost perfect fashion.
27 Reported in the Financial Times, a new artificial intelligence-based software tool called
28 ‘Annie’ aims to relocate migrants to maximise their job prospects through a matching
29 algorithm (Warrell, 2018). The system outperforms humans with early results showing
30 an increase in the refugee employment rate. (Act 1: the setting and the consequence).
31
32 As the article continues, it introduces fear of job losses : “People immediately think,
33 ‘You just want to get rid of all of us?’” and questions of unintended outcomes, bias,
34 transparency become a dominant theme: “there can be a lack of transparency about how
35 final conclusions are drawn”. The final act predictably, identifies governance concerns
36 such as the refugees having the right to know whether the future pattern of their lives is
37 being dictated by a human or a machine.
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51 The benefits of algorithms followed by the superficial relaying of ethical
52 concerns and the pleas for governance is consistent with Gitlin’s definition of framing
53 as ‘persistent patterns of cognition, interpretation and presentation of selection,
54 emphasis and exclusion by which symbol handlers routinely organise discourse’ (Gitlin,
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3 1980, p7, cited in De Vreese (2005)). Characteristics such as identifiable conceptual
4 linguistic features, a recognisable journalistic style, and reliable distinguishing from
5 other types of frames are all visible (De Vreese, 2005).
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10 Thus, the discussion on algorithms permeates the mainstream media. What is
11 uncertain though, is how rich and informed is this debate and what is the nature of the
12 debate? What are the primary concerns? Which moral values are at the core? Does
13 security trump transparency? Unpicking the complexity of how ethics of usage of
14 algorithms is examined is central to this overall discussion and is important for at least
15 three reasons. Firstly, the debate is closely engaged with public policy. For example, the
16 UK Industrial Strategy, places artificial intelligence as a core technology for addressing
17 grand challenges (ageing population and transport infrastructure) (Government, 2018).
18 Secondly, the Select Committee on Artificial Intelligence (AI) reported on key concerns
19 of AI such as the explainability of decisions made through AI algorithms (House of
20 Lords Select Committee, 2018). Thirdly, a joint report by the Royal Society and the British
21 Academy detailed the necessary governance principles and requirements in a future data
22 driven society (Royal Society, 2017). Hence, this paper sets out to explore the public
23 debate from a United Kingdom (UK) perspective. The public debate, enacted in
24 mainstream media, is important because as Diakopoulos (2013, p.2) notes: ‘What we
25 generally lack as a public is clarity about how algorithms exercise their power over us.’
26 And further, ‘new forms of algorithmic power ... are reshaping how social and
27 economic systems work.’ (Kitchin, 2017).
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51 Algorithms and their relevance in modern society has been examined in detail.
52 Gillespie (2014) provides an analytical conceptual map of how algorithms are a key
53 feature of the information ecosystem and the political ramifications arising from the role
54 of algorithms. Such impacts include the choices made about what data is excluded from
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3 an index, how algorithms determine what is relevant to us and as a consequence, how
4 users reshape their practices. Collectively these impacts lead to the creation of a new
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6 knowledge logic that operates on presumptions and adjustments of norms and practices.
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10 With the construction of a new knowledge logic comes the need for additional
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12 governance in the use of algorithms in areas such as explainability, interpretability, and
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14 ethical auditing (Cath, 2018). Algorithms cannot be divorced from the data on which
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16 they are required to operate. Data for example is the basis of ‘hypernudge’ technologies
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18 and consequently requires its own regulation and governance (Yeung, 2017). Principles
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20 of data governance and use have been proposed in a joint report by the Royal Society
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22 and the British Academy that aim to:
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- 25 • protect individual and collective rights and interests
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- 27 • ensure that trade-offs affected by data management and data use are
- 28
- 29 made transparently, accountably and inclusively
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- 31 • seek out good practices and learn from success and failure
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- 33 • enhance existing democratic governance (Royal Society, 2017).
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38 Of note in this report, is the framework that details the multi-dimensional
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40 tradeoffs that explore social and ethical tensions that require choices from society if
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42 data-enabled technologies are to be widely adopted.
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45 Much academic research has been published exploring the ethical dimensions of
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47 algorithms. Various frameworks for conceptualising the key ethical concerns have been
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49 reported (for example, Marda, 2018, Ziewitz, 2016, Pereira, 2018). In 2016, Mittelstadt
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51 et. al. (2016) conducted a systematic review in an attempt to map the ethical problems
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53 prompted by algorithmic decision making. This paper does not therefore present a
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55 background literature to ethics studies of algorithms. Their proposal comprises a
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57 conceptual map that consolidates themes emerging from the literature to a unifying
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3 framework that can serve as an "organising structure based on how algorithms operate"
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5 (Mittelstadt, 2016) and can structure future discussion of how algorithms deal with
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7 ethical issues. Importantly, they recognise the prevalence of "algorithms in the news"
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9 and accordingly, they aim to map the conceptual dimensions of the ethics of usage of
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11 algorithms interpreted along public discourse lines. The framework contributes to more
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13 general approach that aims to translate principles into practices especially for algorithms
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15 embedded in machine learning technologies (Morley et al, 2019). Hence their
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17 framework provides a useful theoretical model for the analysis conducted in this paper.
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19 The framework is described in further detail in section 2.
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24 To do this analysis, this paper reviews articles from UK newspaper articles from
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26 leading newspapers in the UK using the systematic mapping study guidelines developed
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28 in (Kitchenham, 2007, Peterson et al, 2015). The goal is to categorise contributions
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30 from UK newspapers along the ethical dimensions and using the conceptual tool
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32 described by Mittelstadt et al.
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36 The remainder of the paper proceeds as follows. Section 2 outlines the
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38 theoretical framework utilised for the systematic review. Section 3 presents the research
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40 questions and the overarching research framework. Section 4 presents the results and
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42 analysis for the research questions. Section 5 outlines the principal limitations of this
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44 work and finally Section 6 concludes with a summary of the work and directions for
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46 future research.
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50 **2 Theoretical Framework**

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53 Before it is possible to map the ethics debate around algorithms in the newspapers, it is
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55 useful to have key terms defined. Following Knuth's characterisation of the key
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57 properties of an algorithm, there is broad agreement in the computer science discipline
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59 that an algorithm is a set of precisely defined steps for processing an initial set of given
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3 objects (inputs) in order to transform them to a set of outputs such that the sequence of
4 steps terminates in a finite time. As Mittelstadt et. al. points out, algorithms are only of
5 interest in the public media if there is an implementation in a specific technology and
6 the application of the technology is configured for a task (Mittelstadt et al, 2016, pg 2).
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13 The algorithms that are in the news are also mostly concerned with those that make
14 decisions based on complex rules whose detailed workings are difficult to predict or
15 where the actionable insight cannot be explained. Like Mittelstadt et al, algorithms that
16 implement mundane tasks such as "search and replace" are not of interest (Mittelstadt et
17 al, 2016, p3). Nor are abstract descriptions of algorithms of interest. Also interesting is
18 how the notion of algorithm is separated from its host system.
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27 The conceptual mapping tool offered by Mittelstadt et al. is an organising
28 structure for scaffolding discussions of ethical issues. They propose that ethical issues
29 raised by algorithms making decisions, (and implicitly, with little or no human
30 involvement¹), are categorised into six types.
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36 Theories in general can offer routes to analysis and prediction, explanation,
37 prediction and prescription (a description of a method) (Gregor, 2006). Thus the six
38 types of ethical concerns can be seen as a type of theory for classifying dimensions or
39 characteristics and its primary use is for analysis. The six concerns arise from the use of
40 algorithms, how algorithms process data to produce evidence that might motivate an
41 action. Where an action might lead to a potential failure responsibility or accountability
42 means that traceability is also of concern. The six are presented below briefly:
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57 ¹ Mittelstadt et al note that "algorithms augment or replace analysis and decision making by
58 humans" (page 3).
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3 **Inconclusive evidence:** Algorithms in the media draw conclusions from the data
4 that is processed using inferential statistics and/or machine learning techniques. The
5 conclusions are probable and there is uncertainty. Motivation of an action based on
6 probable knowledge is thus an epistemic limitation.
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12 **Inscrutable evidence:** A conclusion based on processed data creates a
13 reasonable expectation that the connection between the data and conclusion should be
14 accessible, intelligible and open to critique. Further, evaluations of the process leading
15 to a conclusion, and whether the evidence produced is misguided, are observer-
16 dependent.
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22 **Misguided Evidence:** Conclusions and actions arising from them² can only be
23 as reliable as the data on which they are based. Informally, computer science
24 traditionally treats this as 'garbage in, garbage out'.
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30 **Unfair outcomes:** Action arising from a conclusion based on processed data
31 should be assessed to criteria or ethical principles broadly understood as "fair" and not
32 discriminatory even if the action is conclusive, scrutable and well-founded.
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37 **Transformative effects:** Algorithms and their prevalence affect how we
38 conceptualise the world in new ways perhaps even modifying people's behaviour and
39 challenging their autonomy. For example, our book buying habits might change through
40 personalisation and nudge technologies³. Personalisation also reduces the diversity of
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49 ² Actions arising from conclusions is added here as a supplementary claim following an earlier
50 statement where Mittlestad et al note that their conceptual map takes into account a given
51 outcome (henceforth conclusion) and the conclusion can then trigger an action. (Mittlestad et
52 al, 2016, page 4).
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57 ³ Personalisation techniques are inherently paradoxical. On one hand, providing only relevant
58 information can aid decision making by reducing information overload. Yet the provision of
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3 information encountered, a pre-condition for autonomy (Van den Hoven and Rooksby, E,
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5 2008).

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7 **Traceability:** Algorithms as software artefacts create further ethical challenges
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9 through design and availability of new technologies as well as complexity associated
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11 with the manipulation of large volumes of personal and other data. An implication is
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13 that the harm caused by software is difficult to debug. The nature of modern distributed
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15 computing techniques, dispersed development activities also means it is “rarely straight
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17 forward to to identify who should be held responsible for the harm caused” (Mittelstadt,
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19 2016, page 5).

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23 A critique of the Mittelstadt et al. paper (Mittelstadt et al., 2016) suggests these
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25 categories provide the bare conceptual structure for categorising and distinguishing
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27 epistemic, ethical and traceability concerns in descriptions of ethical problems related to
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29 the use of (implementations) of algorithms. Significantly, the conceptual map offered by
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31 Mittelstadt et al. (Figure 1 in their paper) does not identify or formally state potential
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33 relationships that might exist between these type structures. Possibly, this could be
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35 based on the approach taken in the systematic review but details of the review step are
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37 not provided and reading of the text does not suggest that it follows along the lines of
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39 accepted models of systematic review processes. However, for the purposes of this
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41 mapping study, the type structure is sufficient with which to explore mapping of the
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43 public debate of algorithms. The next section explains how the theoretical framework is
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45 applied.
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57 the information is subjective and can be coercive in that only institutionally preferred actions
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59 are made possible.
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3 Research Design

Literature reviews in Information Systems (IS) academic research have tended to be *ad hoc* or systemised using either a systematic mapping study approach or a systematic literature review process (Kitchenam, 2004) Mapping studies or *scoping studies* are used to provide an overview of a research area by a process of classification and counting of the outputs (Peterson, 2008, Peterson et al. 2015). Results and analysis are often used to provide a visual map of subject areas and is often coarse grained. More often than not, mapping studies aim to uncover research trends and to provide simple routes to comparison. There are now numerous examples in the literature and the recent mapping study of twenty-eight years of component-based software engineering is cited as an example for the purposes of illustration of how mapping studies reveal research trends (Vale et al., 2016).

In contrast to mapping studies, systematic literature review (SLR) approaches are detailed studies that drill down on a very specific area with the aim of aggregating evidence and are strongly focussed on methods and results. They lend themselves to meta studies of the data, often with the aim of establishing a hypothesis. In a software engineering context, SLRs are more appropriate for meta-analysis of studies that have either quantitative or empirical results (Kitchenam, 2007). For example, Walia and Carver use an SLR approach to review, identify and classify software requirements errors (Walia and Carver, 2009).

Given these differences and the main aim of establishing an overarching map of the ethical concerns of algorithms being debated in the media, it is more appropriate to use a mapping study approach.

3.1 *Data sources*

Two comprehensive data bases were candidates for use (LexisNexis Academic and Factiva) to conduct this media analysis. The former, however had limited functionality from the licensing arrangements available at the author's institution and was primarily used to provide a cross-check facility to ensure that the potentially relevant media were correctly located. Factiva provides a large online collection of newspaper and other media sources such as television and radio transcripts, web blog content and other sources. However, this research focussed only on mainstream newspapers in the UK. Previous social media research has demonstrated that traditional fourth estate media are key brokers of news across all types of social and news networks (Barn et al., 2017).

3.2 *Search strategy*

The Factiva electronic newspaper database was searched for relevant media reports that were published between the dates 21 September 2016 to 21st September 2018. Sources searched were the category Top UK newspapers (both print and online where available). These papers are: Financial Times, The Times (UK), The Daily Telegraph (UK), Daily Mail, The Independent, thesun.co.uk, express.co.uk, The Sunday Telegraph, The Sunday Times and The Guardian (including The Observer. A deliberate focus on the UK was chosen because of several key policy initiatives being enacted by the UK Government. These included the now established Turing Institute (Turing.ac.uk, 2019) the planned Centre for Data Ethics (HM Government, 2019) as well as the 2017 UK industrial strategy with its focus on Artificial Intelligence (HM Government, 2018).

The keywords used were kept deliberately simple to maximise the corpus of data. Keywords used were: "Algorithms AND Ethics". All newspaper articles (print or

online) found were included unless the exclusion criteria identified a duplicate, or the article was not about the type of algorithm described in section 2 earlier.

Figure 1 illustrates the initial search, the screening of the corpus of media reports and the total number of media reports available for analysis.

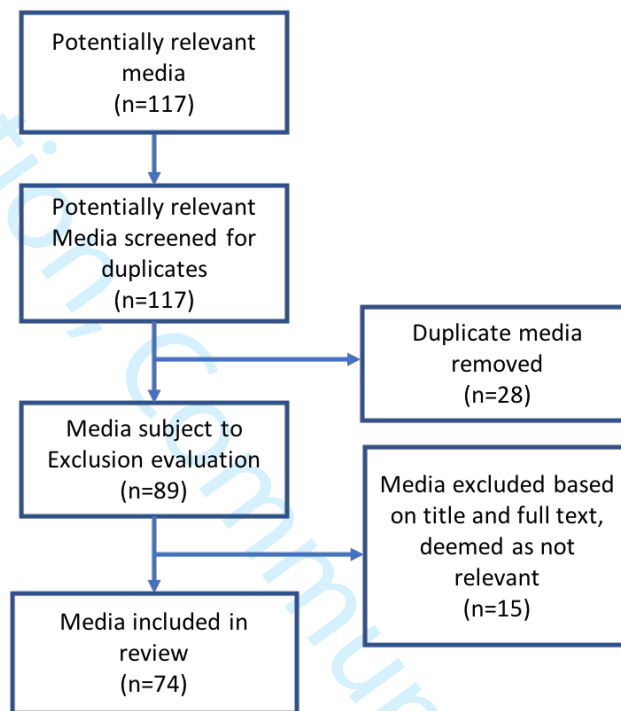


Figure 1: Screening flow diagram

3.3 Themes

Processing of the 74 screened media reports was carried out by close reading of text (Charon, 2017). The media reports are available from the author. The individual articles were allocated to ethical concern categories described in section 2. These categories described by Mittelstadt et. al. help in understanding and formulating a response to the central research question:

What is the nature of the public debate around ethical concerns of the use of algorithms?

Through assignment to the ethical concern categories, it is possible to develop a systematic map of the key concerns being discussed in mainstream media as well as understanding the nature and distribution of the concerns.

Other data / categories were also collected but they are not reported in this paper as they relate to other research questions which are part of a separate study.

Additionally, data generated by Factiva was also available but these data were based on the initial corpus of 117, i.e. before any screening had taken place.

Figure. 2a

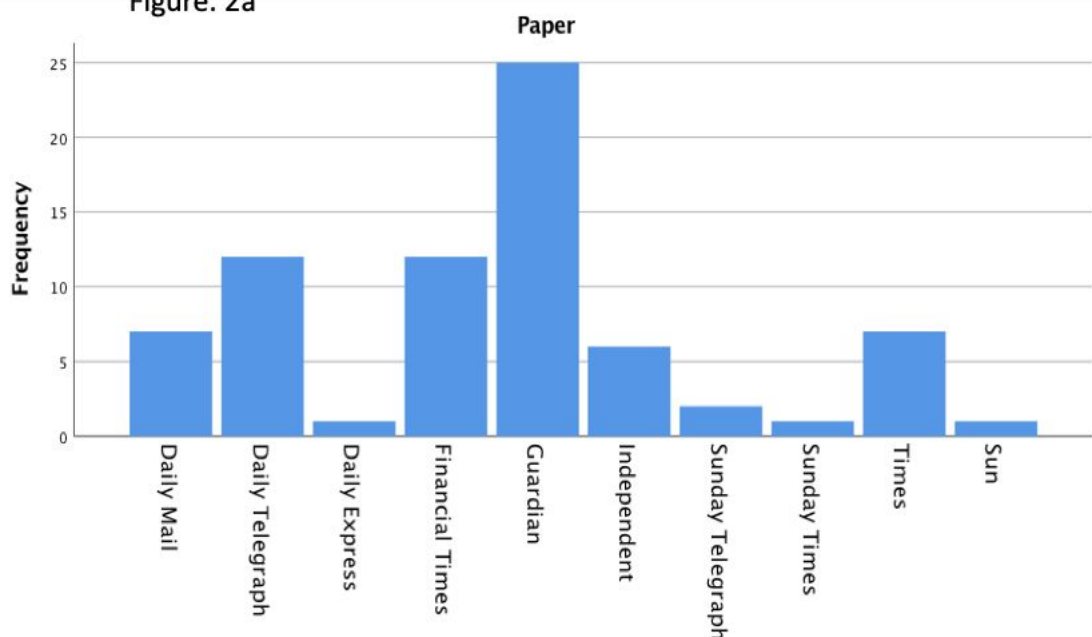


Figure. 2b

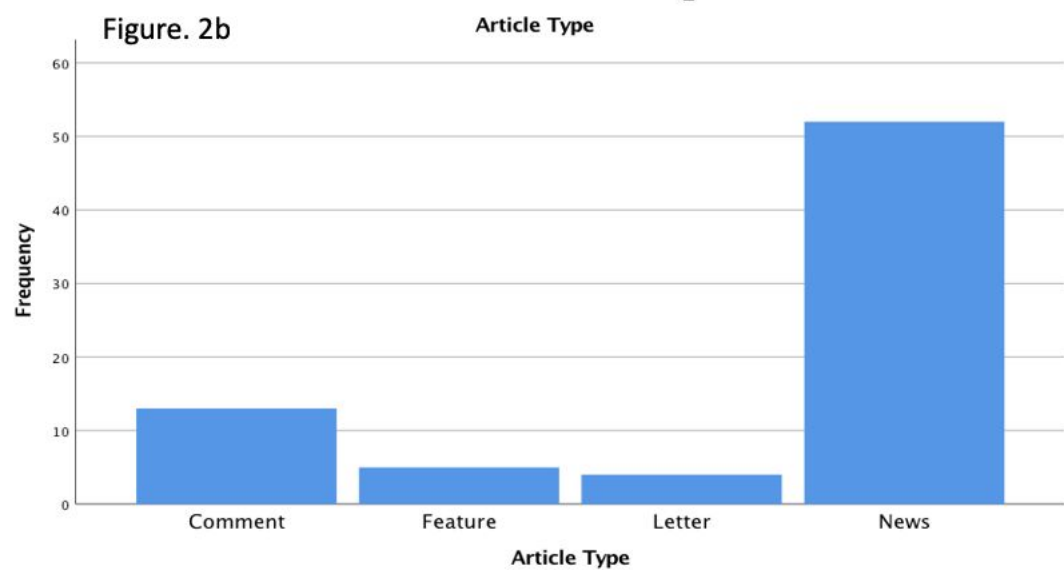


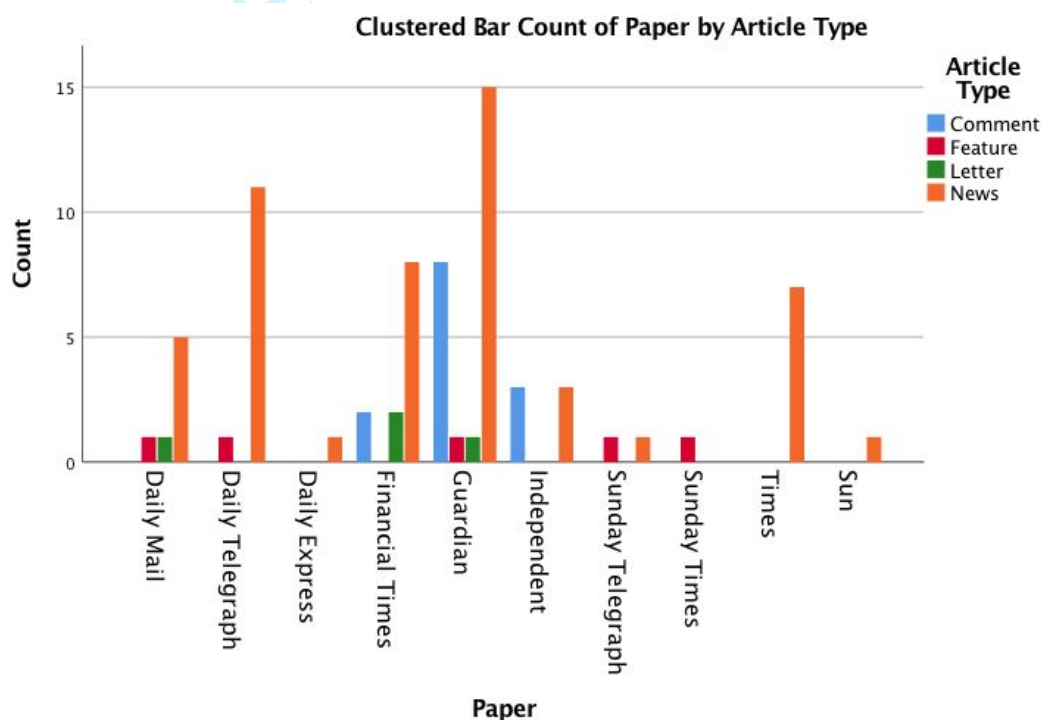
Figure 2: Overall summary

4 Results and Discussion

The media search in Factiva selected 117 reports that mentioned Algorithms and ethics.

Following the screening for duplicates and removal of excluded items, 74 items were cleaned and exported to SPSS (IBM SPSS Statistics for Macintosh, Version 25.0)

for descriptive analysis of the categorisations of ethical concerns. Results from the



descriptive analysis are reported here in line with systematic mapping study requirements.

Figure 3: Types of Articles

4.1 Overall picture

Straightforward descriptive statistics were computed for the key variables, the Newspaper (The Guardian, Financial Times, etc.) and the type of article. During the close reading of the articles, various types of article were identified and are described

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3 briefly here.
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- 6 • Comment (termed opinion in some newspapers), pieces written by experts in the
7 field or those with an interest in the domain.
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- 10 • Feature articles were longer pieces going into depth into key issues and were
11 often based on recent books.
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- 14 • Letters were identified as a usefully separate type given that a letter could be
15 treated as an opinion. They were kept separate as they could represent the lay
16 reader.
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- 19 • News articles are regular articles written by journalists and reported on the use
20 of algorithms at large.
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31 Figure 2 provides a summary of the overall picture of the articles. The nature of
32 newspaper media in the UK does not present any surprises in the overview results given
33 the general position of the newspapers with respect to political leanings (Smith, 2017).
34 The (relatively libertarian) Guardian (ibid) published the most articles and given its
35 approach to attracting commentary and opinion from outside normal journalism also
36 had the most comment pieces (Figure 4B).
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46 **4.2 Ethical concerns**

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49 The conceptual map of ethical concerns proposed by Mittelstadt et. al provides a useful
50 basis for structuring discussions. Following the close reading of the 74 media texts, the
51 texts were assigned to one or more of the ethical concerns. The results in Figure 4
52 present the frequency count of the various concerns along two principal axes. First, (in
53 Figure 4C) we can see that the two dominant themes are Unfair Outcomes (42%)
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(essentially a discussion about potential bias and discrimination), and secondly, Inscrutable Evidence (27%), that is concerns that data and any conclusion are accessible, intelligible and open to critique such that algorithms are not behaving as black boxes. Other concerns were harder to delineate perhaps because unpicking the complexity of say Misguided Evidence or Transformative Effects is not straight forward.

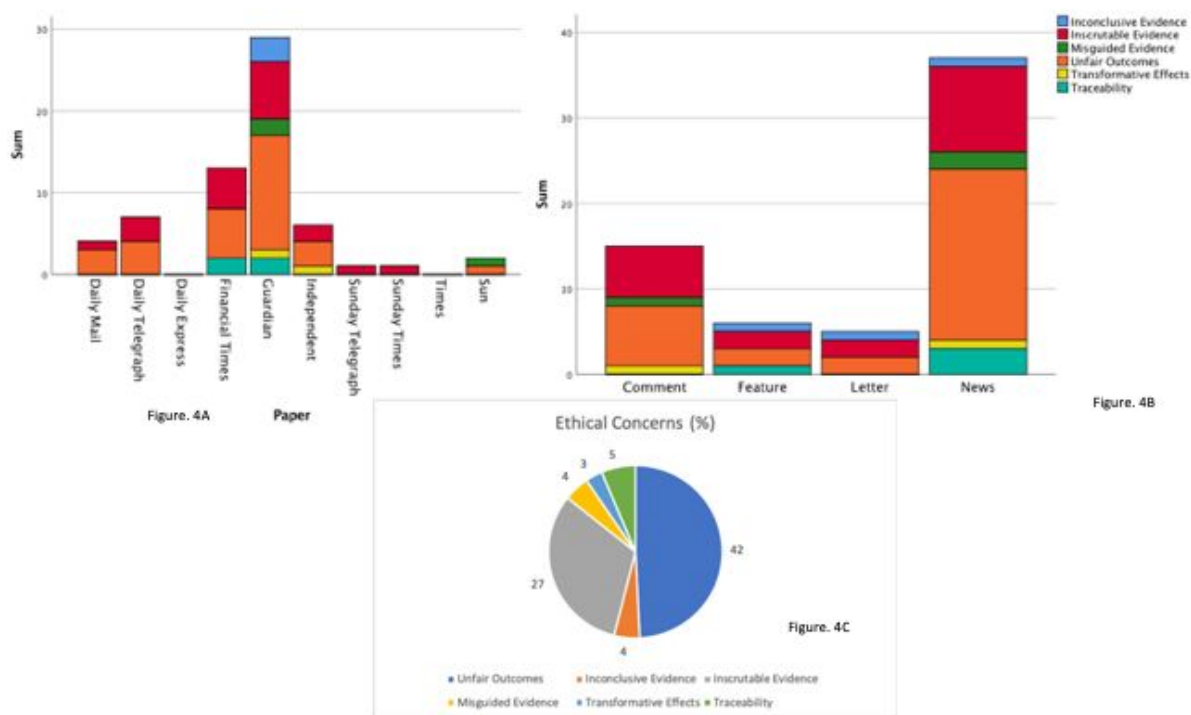


Figure 4: Ethical Concerns

The media reported examples of unfair outcomes along protected characteristics⁴ (e.g. sex, race, religion or belief). Commonly described examples of the use of algorithms often listed the use of predictive analytics in policing and even sentencing (Zarsky, 2016). Most articles however did not go beyond an assertion of a risk of bias.

⁴ It is against the law (Equality Act, 2010) to discriminate against someone because of a protected characteristic.

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3 Indicating that the public debate is lacking sophistication even when reported in quality
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5 newspapers. For example:
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8 "Algorithms are displaying white male bias, and automation is decimating our jobs
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10 we have a lot to lose unless we (women) get involved....If a non-diverse workforce
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12 is creating them, they are more prone to be implanted with unexamined,
13
14 undiscussed, often unconscious assumptions and biases about things such as race,
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16 gender and class." (Bartoletti, 2018).
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18 The debate about algorithms is made more pertinent in the context of the use of
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20 algorithms to process large volumes of data to detect patterns and identify relationships
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22 across vast and distributed data sets (Floridi, 2012). The processes involved use
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24 machine learning techniques, that is, methodologies used to generate models for
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26 prediction and that are capable of defining or modifying decision-making rules
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28 autonomously (Van Otterlo, 2013).
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33 Not all cases are about negative discrimination, one of the most positive sentiments
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35 articles revealed was that the use of machine learning (and indirectly, the algorithms
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37 embedded) helped remove Unfair Outcomes. The use of machine learning based
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39 chatbots to support the recruitment process suggests that neutralising discrimination is
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41 possible:
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45 "...The fact it remains completely neutral with gender, age and race unidentified
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47 keeps things focused on what matters." (Warrell, 2018).
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50 Inscrutable Evidence (27%) featured as the second biggest ethical concern. The
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52 core concern centres around the notion of the algorithm as a black box, where software
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54 does not allow a user to 'look under the bonnet' and allow independent verification of
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56 the inputs that have been entered in order to drive the output. This concern has been
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58 given much emphasis in the select committee report on AI (House of Lords Select
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3 Committee, 2018) which recognised the need for technical transparency to address
4 intelligibility requirements of decisions arising from algorithms. Their conclusion that
5 "full technical transparency is difficult, and possibly even impossible" leads to an
6 alternative proposition of "Explainability" whereby AI systems are developed in such a
7 way that they can explain the information and logic used to arrive at their decisions.
8 This remains an active area of research.
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17 Examples of concerns of inscrutable evidence include the use of predictive
18 analytics tools in social care settings prompted a letter reported in the Guardian:
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22 "...how can government maintain public trust in services and the use of public data,
23 when councils are applying undisclosed algorithms to public data without our
24 knowledge or consent, for unevaluated interventions, to screen their populations
25 for a problem as serious as child maltreatment?" (Gilbert and Pearson, 2018).
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31 Misguided Evidence (4%), the third component of Mittelstadt et al.'s epistemic
32 concerns was reported primarily in the Guardian. For this type of ethical concern, data
33 reliability is central in that conclusions and actions arising from conclusions can only be
34 as reliable as the data in which they are based. The low numbers reported is surprising
35 but perhaps reflects the lack of sophistication of the debate in the public media. The
36 most significant reference to misguided evidence, was in fact, experts critiquing an
37 algorithm that used facial recognition to detect sexual orientation.
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48 "...Todorov and two AI researchers at Google argued that Kosinski's algorithm
49 could have been responding to patterns in people's makeup, beards or glasses, even
50 the angle they held the camera at..." (Lewis, 2018)
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54 The systematic mapping of the ethical concerns and their spread can be seen
55 further in figure 4A and 4B where the quality newspapers and in particular Guardian
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3 reporting most articles about ethical concerns centring on the epistemic issues of
4 algorithms.
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9 **4.3 Relationships between ethical concerns**

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11 The epistemic and normative concerns and the traceability categories offered by
12 Mittelstadt et al. are presented as discrete concepts and formal relationships between
13 these concepts are not offered as part of the concept map. To this author, it is clear that
14 there are complex relationships that might exist between these concepts that are not
15 explicated or are simply alluded to in the narrative text. Hence, to support the mapping
16 study exploration, a further result is presented - a conceptual model for ethical concerns
17 that presents visually the Mittelstadt framework. In this model, the concept map is
18 enhanced with proposed relationships delineating the types and associations between
19 types. The derivation is based on the author's analysis of the Mittelstadt paper. A
20 conceptual model form is chosen as it has the potential to enhance the framework and
21 progress its development to act as a theory. The word "theory" suffers from both an
22 over-use and a reluctance in its use by researchers. Weick comments that most theories
23 that are labeled as theories are actually *approximate* theory in that they go some way to
24 establishing a theory but fall short in some aspect such as: failing to sufficiently
25 articulate relationships between variables/concepts contributing to the theory; or
26 perhaps where ad hoc hypotheses are derived from limited observations (Weick, 1995).
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49 Others have noted that scientific theories demonstrate analogous properties to
50 conceptual models (Evermann and Mistry, 2008):
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54 "In system development, the purpose of the conceptual model is to describe the
55 elements of the domain and their relationships (Mylopoulos, 1992). The conceptual
56 model serves as the basis of understanding and problem solving within the domain.
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It allows analysts to capture and communicate their understanding of a domain and the problems in the domain. In science, it is the role of theories to describe the constituents of the domain, their relationships and their behavior, and to serve as the means by which problems in a domain are specific and solved."

Given this analogy, it appears viable that a conceptual model defining constructs, relations, constraints and possibly behaviours (propositions) could be used as a representative form for describing and further elaborating this Ethical Concerns theory.

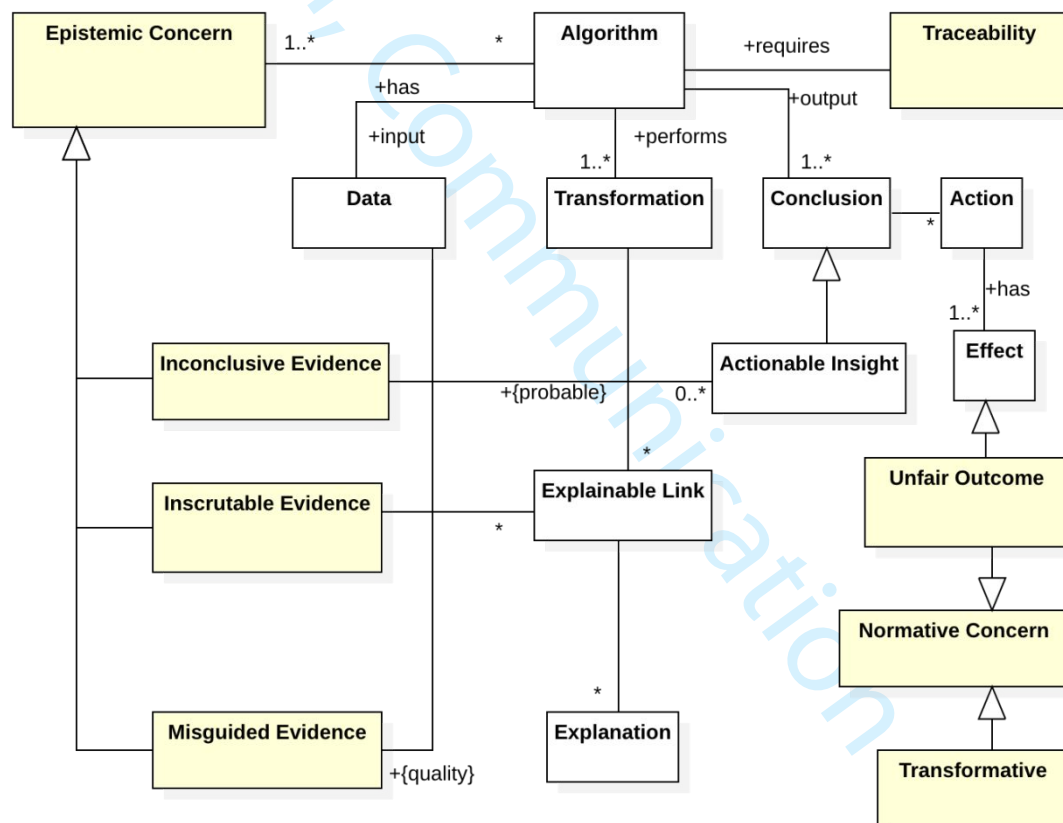


Figure 5. Ethical concerns meta model

As noted above, complex relationships that might exist between these concepts that are not explicated in the original proposal are made visible in Figure 5. The UML conceptual model formalises associations outlined in the descriptive prose in (Mittelstadt et al., 2016). The UML model also allows the formulation of constraints to

1
2
3 add further semantics to the diagram. When a correlation test between the various
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5 concepts was performed three significant relationships were identified. These are shown
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7 in figure 6. The strongest correlation (0.305**) was between Inconclusive Evidence and
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9 Misguided Evidence. In the Mittelstadt et al. model/theory, these concepts both exist
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11 within the Epistemic concerns structure and may therefore be implicitly related. In the
12
13 UML conceptual model shown in Figure 5, these concepts are both subtypes of the
14
15 same supertype (EpistemicConcerns). The data supports that proposal. The two other
16
17 significant correlations are Inconclusive Evidence and Traceability (0.254*) and
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19 Inconclusive Evidence and Unfair Outcomes (0.242*). These cross from the Epistemic
20
21 to Normative Concerns. Here the relationships, while clearly strong, require lengthy
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23 navigation across the model. Both navigation constraints expressed using an English
24
25 language variant of Object Constraint Language (Warmer and Kleppe, 1998) are offered
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27 below.
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34 Context: Inconclusive Evidence: self.conclusion.algorithm.traceability

35 Context: Inconclusive Evidence: self.conclusion.action.unfairoutcome
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39 The complexity created by the navigations across several concepts including
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41 multiplicity constraints offers an explanation as to why ethical concerns centre around
42
43 lack of transparency of decision making and unfair outcomes arising from inconclusive
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45 evidence. Transparency is made challenging because of the number of types involved,
46
47 the relationships traversed, and implication that Actionable Insights are type of
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49 Conclusion and Unfair Outcomes are a type of Effect. The UML model, by breaking
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51 down the relationship into sub-components presents, minimally, a handle on the
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53 complexity.
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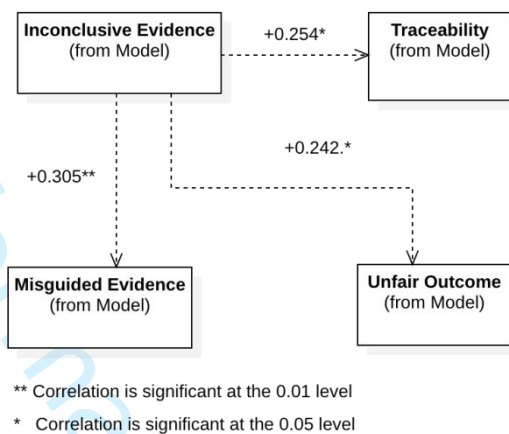


Figure 5 Correlations between ethical concerns

5 Limitations

The principal limitation of the results reported in the study centre around the systematic mapping of the media reports from the main UK national newspapers. Limitations can be presented with respect to construct validity, reliability, internal and external validity as used by Engstrom and Runeson (2011).

Construct validity depends upon the research questions and whether all relevant sources have been taken into consideration. The scope of the mapping was very well defined, the main UK national newspapers as specified by one of the leading news aggregator search databases - Factiva. The key question on construct validity is a discussion on whether the ethical concerns framework used in this paper is sufficiently robust. There is a reliance and assumption that peer review, academic rigour and subsequent citation of the Mittelstadt et al. model offers some confidence regarding suitable construct validity. Categorisation of media with respect to each ethical concern is dependent upon close reading the text, but it is accepted that errors in allocating categories may be possible.

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3 Reliability concerns the arrival of a similar conclusion through repeating the
4 study by another researcher. The protocol definition phase and aspects of the execution
5 phase are repeatable by other researchers. The Mittlstaad et al. framework offers a base
6 benchmark for using pre-defined categories offering a significant opportunity for
7 building in reliability for future studies.
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12 External validity is about generalisation about the results from this study. This
13 paper does not draw any conclusions about mapping studies in general hence, external
14 validity threats are not applicable here.
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19 The scope of the mapping study could have been bigger. For example, a longer
20 period could have been chosen. It is noted however, that even this period from 21
21 September 2016 to 21 September 2018 saw one of the biggest issues around the use of
22 algorithms and ethical concerns explode with the news story concerning Cambridge
23 Analytica and the harvesting of 50 million Facebook profiles for political analysis
24 purposes (Graham-Harrison and Cadwalladr, 2018).
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36 **6 Conclusion**

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38 Algorithms are increasingly in the public eye through articles in the mainstream media,
39 the so-called fourth estate. Their role, especially, in the context of artificial intelligence /
40 machine learning is in decision making in all aspects of society. The key debate of the
41 role of algorithms centres around ethical concerns such as the transparency of decision
42 making and unfair outcomes arising from decision making. The work described here has
43 made three principal contributions around this debate. Firstly, a systematic mapping
44 study of leading UK national newspapers for one year has been conducted and a map of
45 the key ethical concerns being debated has been produced. Secondly, the paper has
46 provided experimental evaluation of a significant framework for evaluating ethical
47 concerns. Finally, the research has contributed an initial UML conceptual model of this
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framework thereby adding further depth and texture to the Mittelstadt et al. conceptual framework. In particular, the conceptual model exposes dependencies between the two principal elements of the original framework, the Epistemic and Normative concerns.

While the presentation of algorithm articles follow a media framing model, future research could explore more precisely the defining characteristics of algorithm framing in the media and therefore allow for more substantive critique of such articles.

Future studies could adopt the core approach taken in this paper by developing further mapping studies and perhaps systematic literature reviews of both public media and academic literature to explore specific relationships exposed in the UML conceptual model.

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