

Modeling the Sacred: Considerations when Using Religious Data in Natural Language Processing

Anonymous ACL submission

Abstract

This position paper concerns the use of religious data in Natural Language Processing (NLP), which is of special interest to the Ethics of NLP. Religious texts are expressions of values and cultural practices that relate to deeply held convictions, and machine learned models have a propensity to reproduce cultural values and biases encoded in their training data. Furthermore, translations of religious texts are increasingly being used by researchers and developers, especially when language data is scarce. This repurposes the translations from their original uses and motivations, which often involves attracting new followers. With these in mind, this paper discusses the considerations of using religious texts for the development of language technologies, including concerns around cultural rights.

1 Introduction

The Association for Computational Linguistics (ACL) is a secular institution. Its constitution, resolutions and policies make no mention of religion other than forbidding harassment on the basis of religion.¹ Nevertheless the Christian Bible and the Islamic Quran² are increasingly being used in the scientific and professional activities of ACL, as measured by papers published in the ACL Anthology (Figure 1). Some of the reasons that NLP researchers use the Bible are aptly expressed by Resnik et al. (1999). The Bible is the world’s most translated book, with translations in over 2,000 languages, and often multiple translations per language. Furthermore, great care is taken with the translations, so from an NLP perspective data quality is high. It is often easily available in electronic form, and is in the public domain, hence free to

¹www.aclweb.org, accessed September 2023

²This paper follows several style guides in using “Quran”, although mentions of the alternate Latinization “Koran” are also considered in the corpus studies we report on.

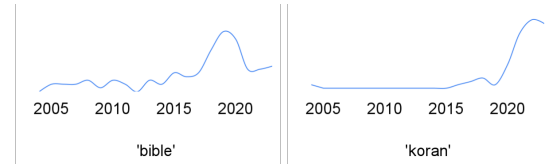


Figure 1: Trend sparklines for counts of papers in the ACL Anthology mentioning ‘bible’ or ‘koran’ (see §3).

use. It has a standard structure which allows parallel alignment verse-by-verse. (And some of these reasons, or similar ones, hold of the Quran too.) For these reasons, as recently as 2006 it was said to be “perhaps surprising that the Bible has not been more widely used as a multilingual corpus by the computational linguistics and information retrieval community” (Chew et al., 2006).

Despite the increasing use of sacred texts in NLP, the ethical considerations around such use does seem not to have received prior attention (other than a brief mention by Mager et al. (2023)). This position paper contends that responsible secularism demands engaging with the ethical considerations of the use of sacred texts. If one should never speak of religion in polite company, then perhaps ACL forums should be less polite. In Section 2, we provide relevant background, and summarize a debate within the field of academic linguistics concerning its disciplinary relationship with missionary linguistics. In Section 3, we present a study of papers in the ACL Anthology mentioning sacred texts, as well as four recent case studies that illustrate ways in which NLP research uses sacred texts. Section 4 discusses a range of ethical considerations when using sacred texts in NLP. In doing so, we consider a range of approaches to the topic, including ethical theories, Indigenous perspectives,³ human rights, and the AI principles commonly espoused by institutions. Our goal in doing so is not to eval-

³Following several style guides, we capitalize the first letter of “Indigenous”. See, e.g., <https://www.sapiens.org/language/capitalize-indigenous/>.

Religion	Sacred texts	Est. 2020 population	Proselytizing
Christianity	Bible, New Testament, Old Testament	2,382,750,000	Yes
Islam	Quran (alt. spellings include Koran), Old Testament	1,907,110,000	Yes
Hinduism	Vedas, Upanishads, Puranas	1,161,440,000	No
Buddhism	Tripitaka, Mahayana Sutras, Tibetan Book of the Dead	506,990,000	Yes
Traditional Chinese Religion	Zhuangzi, Tao-te Ching, Daozang	310,000,000	No
Judaism	Talmod, Torah, Tanakh, Old Testament	14,660,000	No

Table 1: Some major world religions and their texts. Population estimates are from the US-based Pew Research Center (www.pewresearch.org), which conducts demographic and other research.

uate past NLP projects that use religious texts, but rather to encourage more reflecting in and on future work. Based on these considerations, we then make some recommendations for the NLP community in Section 5, concerning cultural standpoints, cultural knowledge gaps, and power dynamics between global and marginalized cultures.

2 Background

2.1 Religion

Precisely defining what constitutes a religion might be notoriously difficult (see e.g., [Spiro, 2013](#); [Neville, 2018](#)), and lies beyond the scope of this paper. Common properties of religions center around giving meaning to existence, and include: a) moral values concerning which actions are right or wrong, b) spiritual beliefs, including what happens after death, c) theistic beliefs about gods, or spirit beings, d) rituals around birth, initiation, obtaining adulthood, marriage, and death, e) stories and mythologies concerning topics such as the origin of the world, the origin of humans, etc, f) kinship systems and marriage practices, g) artistic practices, including songs, dances, and visual arts, h) significant locations, including buildings, sites, and homelands; and in some cases i) a language which plays a special role. These are all closely related to questions concerning values: moral values, spiritual values, cultural values, aesthetic values, historic values, and even linguistic values.

Like languages, religions may exhibit regional variation and incorporate local practices, so thinking of them as discrete entities may be somewhat misleading. Some widely cited estimates put the number of worldwide religions at several thousand, although these claims are disputed. What seems more certain is that the imminent extinction of many Indigenous languages will accompany an “impending loss of so many religions and world-views” ([Harrison, 2007](#), p. 153). Acknowledging these challenges, we nevertheless provide a sum-

mary of some of the world’s most populous religions in Table 1. Surveying or defining each of these religions is beyond the scope of this paper. Also not included here are the various sects and branches within each religion (e.g., Catholicism), nor texts which might be important only to specific branches (e.g., *The Book of Mormon*).

One important distinction is that between proselytizing and non-proselytizing religions. The former attempt to convert new populations, whereas the latter do not. The former are more intricately related to historical practices of colonialism—especially in Africa, the Americas, Asia, and the Pacific—and hence also to neocolonial legacies. Some religions hold that a certain language is privileged for communicating sacred texts to the faithful, while on the other hand Protestant Christianity exemplifies a commitment to communicating in vernacular languages. (Article XXIV of the Articles of Religion of the Anglican Church calls for “such a Tongue as the people understandeth”.)

2.2 The Academy and Bible Translation

The September 2009 issue of the journal *Language* has a special feature of 5 articles by anthropologists and linguists concerning the relationships between the US-based Bible translation organization SIL International (SIL) and academic linguistics. In this issue, [Dobrin and Good \(2009\)](#) explore how academic linguists have at times become reliant on, and benefitted from, the technological infrastructures of SIL, in part because creating and maintaining these infrastructures has not been valued by the academy. This “partnership of convenience” causes tensions between differing objectives, and raises questions about what kind of relationships secular research institutions should have with organizations with very different agendas. These practices presage similar ways in which some areas of NLP research have become reliant on Bible translations. Many linguists and NLP practition-

ers working on Indigenous languages see their research as addressing issues of human rights and cultural extinction. However, as Dobrin and Good point out, languages which are most endangered are least likely to receive SIL’s attention. Handman (2009) draws attention to how SIL ideology separates linguistic identity from religious identity, differing from UNESCO’s position that sustaining endangered languages entails sustaining cultural worldviews, knowledge systems, and identity practices. Epps and Ladley (2009) argue that evangelical success entails the displacement or transformation of traditional beliefs, often leading to social upheaval, and argues that the academy has a moral interest in supporting local self-determination which is at odds with evangelical agendas.

3 The Use of Sacred Texts in NLP

In this section we demonstrate that the use of religious data is common in the field of NLP when using machine learning to train models. We focus primarily on the publications of the NLP research community, as represented by the searchable Anthology of the Association of Computational Linguistics (a.k.a., the ACL Anthology) (Bird et al., 2008).⁴ We acknowledge that this corpus might not be representative of the entire field of NLP, e.g., missing relevant work such as (Chandra and Ranjan, 2022; Bashir et al., 2023), and NLP projects in industry might not be well represented in the ACL Anthology. Another limitation is that we exclude publications in languages other than English.

3.1 Sacred Texts in the ACL Anthology

The number of ACL Anthology entries for each of the texts listed in Table 1 are shown in Table 2. Thousands of papers in the ACL Anthology seem to use religious texts. There is a strong bias towards the texts of the monotheistic Abrahamic religions of Judaism, Christianity, and Islam. All three originate in the Middle East, with original versions of their texts in Hebrew, Ancient Greek, and Classical Arabic. However Christian texts have been widely distributed in European languages following the Protestant Reformation, and the global spread of Christian influence is associated with European history and the colonial and missionary practices of European cultures.

To understand *when* sacred texts have been used by the NLP community, we examined the year of

Search term/phrase	Results [min, max]
bible	[1920, 3890]]
quran	[291, 547]
new testament	[294, 294]
koran	[131, 248]
old testament	[73, 206]
torah	[25, 153]
talmud	[21, 22]
vedas	[22, 51]
tripitaka	[7, 7]
upanishads	[6, 6]
mahayana sutras	[4, 4]
tanakh	[3, 4]
zhuangzi	[3, 3]
puranas	[3, 3]
tibetan book of the dead	[0, 0]
tao-te ching	[0, 0]
daoziang	[0, 0]

Table 2: Number of search results for religious texts in the ACL Anthology on August 10, 2023. Since ACL Anthology search result counts are non-deterministic, we report the min and max of 10 searches for each term.

publication of the first 100 results (sorting by relevance, not by date) on the ACL Anthology for each of the terms ‘bible’ and ‘koran’. (The ACL Anthology search interface only made these available.) We omitted 12 ‘bible’ search results from our analysis due to either: not being in English (2); not being research papers (6, e.g., book reviews, invited talks, or proceedings); or being duplicates (4). We omitted 36 ‘koran’ search results from our analysis due to either: not being research papers (16); being false positive search results (13, e.g., 10 had typos for ‘Korean’); or only using the word ‘koran’ in references (3), footnotes (2) or content generated by a model (2). We manually coded each paper for its year of publication, NLP application domain and which languages it covered. 150 of the resulting 152 papers spanned from 2004 to July 2023, and of these we see increasing use of the Bible and the Koran in NLP research over time (see Figure 1 in Section 1⁵), including over 60% of papers being published between 2019 and 2023 alone.

To understand *how* sacred texts are used in NLP, we analyzed the 88 papers mentioning the Bible described above. A variety of application domains were represented, including sentiment analysis, named entity tasks, CLIR, patronizing language detection, and various morpho-syntactic analysis tasks. However the most common applica-

⁵Since our 2023 sample was limited to Jan-July, paper counts for 2023 are multiplied by 12/7 when creating sparklines, to make them comparable with previous years.

⁴www.aclanthology.org, accessed August 2023.

tion domain was machine translation (22%), while many (18%) introduced a new corpus or lexical resource. Three papers were concerned specifically with Bible translation, four with literary analysis of the Bible, and five with language modeling or pre-training. There seems to be a recent trend towards papers handling very large numbers of languages, with ten papers since 2014 in our sample handling over 500 languages. 48% of the 88 papers concerned one or more Indigenous languages.⁶ The 64 ACL papers mentioning ‘koran’ were less varied and overwhelmingly used verses included in machine translation evaluation datasets.

3.2 Four Contemporary Case Studies

To complement the broad analysis of the previous section, we also report here in more detail on some recent noteworthy NLP papers. These illustrate in more detail ways in which the NLP community is encountering and using religious texts.

Our first example is a paper which was uploaded to arXiv, a popular online archive for computer science papers, in May 2023 (Pratap et al., 2023). This work aims to improve Speech Recognition and Text To Speech Synthesis for over a thousand languages. They train their model using translations of the New Testament, as well as audio of readings of those translations, obtained from Faith Comes by Hearing (faithcomesbyhearing.com), goto.bible and bible.com. They also use spoken recordings in many languages, without paired texts, of Bible stories, evangelistic messages, scripture readings, and songs, obtained from Global Recordings Network (globalrecordings.net), whose mission is to communicate “the Good News of Jesus Christ” via a strategy of recording, distribution, and promotion.

Our second example was awarded the ACL Area Chair Award for best Multilingualism and Cross-Lingual NLP paper in July 2023 (ImaniGooghari et al., 2023). This work aims to scale language models to 500 languages. They “crawl or download” data from 150 sources, including religious texts and observe a “higher proportion of religious data” compared to previous comparable work. Parallel verses from Bible translations are used for model training and testing, and performance is reported for Sentence Retrieval from the Bible.

⁶These categorizations were done by the author, taking into account historical and social context, however an ideal approach might engage with language communities to understand whether they consider themselves Indigenous.

Our third example concerns JW300 (Agić and Vulić, 2019), a dataset of around 100k sentences in each of 300 languages crawled from jw.org, a website run by the US-based Jehovah’s Witnesses, a Christian denomination. A majority of the texts come from the Jehovah’s Witnesses’ magazines *Awake!* and *Watchtower*. Released as a corpus in 2019, JW300 has been cited over 150 times as of August 2023. The African grassroots open-source NLP project MASAKHANE (masakhane.io) had been using JW300 to train Machine Translation models, until receiving legal advice in 2023 that this was breaching copyright. A subsequent request by MASAKHANE to the Jehovah’s Witnesses for permission to use the data was declined.⁷

Our final recent example concerns the release of MADLAD-400, a new text dataset containing 3T tokens in 419 languages (Kudugunta et al., 2023). It uses 2022 snapshots of the CommonCrawl web crawl (commoncrawl.org) and the paper was uploaded to arXiv in September 2023. Auditing of a preliminary version of the dataset, spanning 498 languages, revealed that for 141 languages there were “significant amounts” of Bible data. Significant amounts of Jehovah’s Witnesses data was also found for 37 languages, and of Church of Jesus Christ of the Latter Day Saints (LDS) data for 2 languages. (No Quran data was reported to be found in significant amounts.)

4 Considerations

Having demonstrated above that religious texts have been used in thousands of NLP papers, we now discuss some of the ethical considerations. Our goal is not to critique the work discussed in the previous section, but rather to provide a toolbox for assisting critical thinking in the future. Following a call for NLP researchers to focus their ethical considerations on power relations between technologists and communities (Blodgett et al., 2020)—and cognisant that we are in the International Decade of Indigenous Languages (2022–2032)—we give special attention to relationships between local Indigenous communities and global projects, for which large power disparities exist and for which “the asymmetry of power is the cause of domination” (Mager et al., 2023).

⁷See, e.g., <https://walledculture.org/a-blatant-no-from-a-copyright-holder-stops-vital-ling> or <https://www.youtube.com/watch?v=mbkuRZkg1RY>.

4.1 On Cultural Relativism

One possible objection to the claim that such considerations are needed is based on cultural relativism. Such an objection would argue that by using sacred texts for an extended period, the ACL community has demonstrated its acceptance of such practices. That is, such practices should be judged as acceptable by the norms of the ACL community.

We counter that such an objection would be stronger if the ACL community both had a stronger history of reflexive practices, and was more culturally diverse. Compared to many other disciplines, we find that ACL’s interest in ethics to be relatively recent. The first Workshop on Ethics in NLP was held in 2017 (Hovy et al., 2017). “Ethics and NLP” was not a possible submission topic for the ACL conference until 2020. The ACL adopts the ACM Code of Ethics,⁸ a general code for computing professionals which makes no mention of working with cultural data such as language. In 2022, a Responsible Research Checklist was introduced. In 2023, seemingly for the first time, the ACL encourages (but does not require) an ethics statement. Unlike some other disciplines, positionality statements are rare in ACL papers, with “researcher positionality” and “author positionality” each having only a single (and recent) result in the ACL Anthology as of September 2023. We also find that the ACL community does not represent the diversity of the world’s local languages and religions; in fact disparities in sources of ACL publications might be increasing (Rungta et al., 2022).

4.2 The Veil of Religious Ignorance

A useful starting premise may be that ethical consideration of the papers in the journals and proceedings of the ACL community should not be biased against nor towards any religions. Rawls’ Veil of Ignorance suggests a guide (Rawls, 1971): that one’s consideration of the use of sacred texts within NLP should proceed as if one were in ignorance which religion (if any) we each belong to. For example, how would I feel about NLP’s reliance on texts from major religions if my own culture and religion might be marginalized and endangered?

4.3 Etic NLP and Emic NLP

Although the Veil of Ignorance might seem useful in theory, our ability in practice to truly avoid

⁸<https://www.aclweb.org/portal/content/acl-code-ethics>

being informed by our own cultural backgrounds and affiliations is highly questionable. We must instead accept that we have cultural standpoints. The emic/etic distinction originated in linguistics in the 1950s for describing different standpoints for language research (Mostowlansky and Rota, 2020). *Emic* is commonly used to describe research on a culture from the perspective of people of that culture. This contrasts with *etic* research, which takes an outsiders perspective. We propose it is useful to talk of *etic NLP* and *emic NLP*, according to whether the language technology is for our own linguistic cultures or those of others. Similarly, when NLP handles religious texts, we can distinguish research problems and applications which are within the researcher’s own religious context, from those applications which impact those having other religious beliefs (e.g., translation for the purpose of proselytizing).

The more different that a culture is from our own, the less we should expect to understand the ethical considerations that those of that culture may have around language technologies. Of special consideration here are Indigenous cultures, which perhaps diverge the greatest from the majority global cultures. These are grappling with historic and current marginalization which often places their worldviews, languages and religions at risk of extinction, and their voices are among the least likely to be represented in ACL’s prestigious conference presentations, plenaries, panels, and journals.

4.4 A Range of Lenses

In the previous subsections, we have already begun to lay the groundwork for our position that the ACL community should adopt an acknowledge our own individual cultural standpoints, ignorances of other cultures, and relationships to global and marginalized cultures. With this in mind, we now consider the use of sacred texts in NLP from a range of moral and sociological lenses.

The term **consequentialism** refers to a family of normative theories which emphasize the importance of considering the consequences of actions. An example is **utilitarianism**, which holds that the best action is one that maximizes wellbeing and minimizes suffering. It has been argued that AI research often implicitly adopts a utilitarian lens (e.g., Hutchinson et al., 2022). One challenge with applying these theories in an NLP research context is that in practice research activities are often far

414 removed from applications, and any eventual path
415 between the two can be unknowable or uncertain.
416 Even if the impacts on users of NLP applications
417 can be known, challenges arise in calculating ag-
418 gregate benefits across disparate stakeholders with
419 different objectives. Perhaps the easiest outcomes
420 to reason about are the impacts of NLP research
421 and applications on the lives of the researchers
422 and developers themselves, since research papers
423 and software bring rewards from within the NLP
424 community itself, in the forms of kudos, citations,
425 career progression, etc. How can these be weighed
426 against the possible risks of dignitive harms, e.g.,
427 if believers feel that NLP systems operating with
428 a non-trivial error rate are offensively trivialising
429 sacred texts? If copyrighted parallel texts are pur-
430 chased from a proselytizing organization, what
431 are the downstream impacts of the organization
432 re-investing those monies into efforts to convert?

433 The term **deontology** refers to a normative the-
434 ory which posits that there are rules or principles
435 which determine the rightness of wrongness of ac-
436 tions, rather than the consequences. Within an
437 NLP context, this might lead to a focus of on the
438 upstream actions such as sourcing of data, rather
439 than downstream actions such as usage of NLP ap-
440 plications. Within the context of sacred texts and
441 their translations, this might lead to questions such
442 as: Were the right people involved in the creation
443 of a dataset of sacred texts? Did they have the right
444 roles and relationships from within the perspec-
445 tive of followers of the religion? Do translators
446 of sacred texts have the right specialist translation
447 skills and cultural knowledge? Were the translators
448 paid fairly? Generally, was the dataset collected
449 in a manner aligned with best research practices,
450 e.g., as operationalized by research ethics boards?
451 [Prabhumoye et al. \(2020\)](#) discuss the importance of
452 informed consent for deontological approaches to
453 NLP ethics, and community-level consent might be
454 an appropriate lens for thinking about communities
455 of religious practice.

456 We use the term **AI Principles** here to refer
457 broadly to the hundreds of sets of principles for
458 responsible and ethical AI that have been released
459 by companies and governments in recent years. Al-
460 though all unique in their own way, the sets of prin-
461 ciples also have many facets in common ([Floridi
462 and Cows, 2022](#)), some of which we discuss here.
463 From a *safety* perspective, could the use of sacred
464 texts be used in misinformation or disinformation?

465 For example, could an NLP system trained on sa-
466 cred texts be used in such a way that, deliberately
467 or accidentally, leads to false beliefs about a reli-
468 gion? From a *privacy* perspective, when dealing
469 with speech recordings of sacred texts, does the in-
470 herently identifiability of human voices introduce
471 concerns which can't be mitigated? From a *bias
472 and fairness* perspective, what kinds of cultural
473 biases and encodings of values are present in the
474 sacred texts and likely to be reproduced by sys-
475 tems trained on those texts? Given that sacred
476 texts are poor representations of other linguistic do-
477 mains (see discussion in e.g., [Mayhew et al., 2017](#);
478 [Adelani et al., 2021](#)), what kinds of system biases
479 are likely to result when systems trained on sacred
480 texts are used in other domains? From an *account-
481 ability* perspective, who is accountable for inappro-
482 priate behaviours of systems trained on religious
483 texts? What are their accountabilities are there
484 to followers of religions if systems produce offen-
485 sive religious language or misinformation? From
486 a *transparency* perspective, have datasets incorpo-
487 rating sacred texts followed best practices around
488 dataset documentation (e.g. [Bender and Friedman,
489 2018](#); [Gebru et al., 2021](#); [Pushkarna et al., 2022](#))?

490 One topic in the **sociology of technology** is con-
491 cerned with how power is distributed. As [Win-
492 ner \(2017\)](#) has argued, artefacts are political in
493 the sense that they make it easier for some people
494 to do some things, and perhaps more difficult for
495 other people to do other things. Languages, like
496 language technologies, have politics. Following
497 calls by [Blodgett et al. \(2020\)](#), we might focus our
498 considerations on which NLP tools trained on sa-
499 cred texts serve to re-arrange power, by considering
500 which actions are encouraged or discouraged and
501 by who? And we can consider the etic/emic dis-
502 tinction (§4.3): how are those within or outside the
503 religious community empowered or disempowered,
504 compared to others? For example, accurate ma-
505 chine translation into a language might inhibit the
506 need for paying human translators from a marginal-
507 ized community.

508 As discussed in Section 3, many NLP papers
509 using the Bible use translations of its texts into
510 Indigenous languages. There are many **Indige-
511 nous lenses** on responsible research, technology
512 and data practices, as well as guidelines for non-
513 Indigenous researchers working with Indigenous
514 data (e.g., [Carroll et al., 2020](#); [National Health
515 and Medical Research Council \(Australia\), 2018](#);

Taiuru, 2021). However, despite the plurality, some common themes emerge regarding concerns when working with Indigenous communities and their cultural data, including linguistic data. One theme is around *relationships*: Does the researcher have a relationship with the Indigenous community which involves generosity, reciprocity, humility, responsibility, obligations, and care? Another is around *benefits*: Does the Indigenous community benefit, collectively and individually, from the NLP project? Are benefits shared equitably? Is the community empowered by the project, and are their capabilities and capacities improved? Are possible harms appropriately mitigated? A third theme is around *culture*: Does the project respect the spirit and integrity for all facets of the Indigenous culture? Are cultural concerns around secrecy and privacy respected? Does the NLP project help to maintain Indigenous culture and connections to culture, assisting with continuity and mitigating threats to extinction? A fourth theme is around *control*: With concerns around Indigenous data sovereignty in mind, are Indigenous people in control of their cultural data? Can projects and their outcomes be contested by the Indigenous community?

Further considerations around **international laws and human rights** concern NLP’s use of sacred texts translated into Indigenous languages. As tools of colonizing projects, such translations have been described as a “well documented example of the non-ethical misuse of translation” (Mager et al., 2023). Kenyan human rights scholar Makua Mutua describes at length what he calls the “basic contradictions” between proselytizing religions and Indigenous cultures (Mutua, 2004). Observing that religion is woven into every aspect of social and cultural life in Indigenous cultures, including dances, ceremonies, rites, and marriage practices, Professor Mutua argues that the meeting of such cultures with proselytizing Christian and Islam faiths amounts to “cultural genocide”. In some cases, this characterization seems valid, and also true of Indigenous encounters with other dominating ideologies, for example the brutal repression of Indigenous religions by Soviet Russia led to extermination of religions without a trace (Harrison, 2007, p. 152). In other cases, “cultural genocide” may be too strong a phrase for the complex realities of how local cultures respond syncretically to proselytizing cultures. For example, Australian historian Laura Rademaker describes how on Groote

Eylandt, an island in remote Northern Australia, the Indigenous Anindilyakwa people reinterpreted Christianity of the missions in their own ways leading to a “hybridisation of cultures” (Rademaker, 2014). Prof. Mutua argues that the right to freedom of religious belief cannot be considered to exist in a “level playing field” in which local cultures can compete with global ones. Rather, the contexts of cultural invasion unfairly privilege global religions, including missionaries making access to education and health services conditional on the “salvation” of “infidels”. This echoes arguments by legal scholars that the power and sovereignty dynamics between source and target cultures constitute important factors between “proper” and “improper” proselytism (Stahnke, 1999). The Human Rights Committee has acknowledged that the cultural rights protected under Article 27 depend on the ability of a minority group “to maintain its culture, language or religion”.⁹ As such, Prof. Mutua argues that the (then Draft) UN Declaration on the Rights of Indigenous People appears to prohibit proselytizing by agents external to the the Indigenous culture in order to create space for Indigenous peoples to maintain their cultures amidst external threats. It follows that one consideration for NLP is whether using Indigenous language translations developed by proselytizing projects constitutes complicity with, and promotion of, projects which might violate international human rights to maintain Indigenous cultures.

5 Recommendations

We take the position that the ACL should strive to be aware of risks of harms to religious communities that NLP research may cause or be complicit with. We suggest that discussions of **Ethical Considerations** should be more common in ACL papers using religious data, and hope the previous section is useful for these. We now discuss how the NLP community might better engage with its cultural standpoints with respect to religious data, and relationships to marginalized religious communities.

Do No Harm. Rogers et al. (2021) provide a useful checklist for responsible data use in NLP, advocating for a principle of “Do No Harm”, which includes considering the potential for misuse. Focusing on Indigenous language technology, Schwartz (2022) also argues that the most important concern

⁹General Comment No23, UN Doc. IC-CPR/C/21/Rev.1/Add.5 (1994), para. 6.2.

is not causing harm. Due to their relationship to deeply held worldviews, we suggest that the use of sacred texts should also prioritize this principle. This entails understanding the harms from the perspectives of the religious communities impacted by the work, for example, a system for automatically translating texts of a global religion into a local Indigenous language should consider possible impacts on those Indigenous communities.

Etic NLP projects. When should an NLP practitioner tread more carefully? We suggest that when handling sacred data from other cultures and religions there will be more morally consequential risks, since the epistemic and axiological uncertainties are greater, if not the aleatoric ones too.

Positionality statements. Linguistic positionality statements have been recommended by researchers who are aware of how different priorities and agendas between researchers and language communities can impact projects (e.g., Rolland et al., 2023; Cormier, 2018), and are exemplified in NLP by Ghosh and Caliskan (2023).¹⁰ Similarly, we suggest that religion positionality statements, for NLP research working with religious data, can also provide a useful signal for the NLP community concerning agendas. For example, hypothetically, would a group of non-Muslim researchers have the Muslim community’s concerns at heart in developing an automated exegesis system for the Quran?

Transparency. As discussed in Sections 2 and 3, both linguistics and NLP have become dependent to varying degrees on datasets developed by missionary projects. Indeed, even the International Standards Organization’s codes for languages, used by many NLP projects, are maintained by SIL (Doarin and Good, 2009; Morey et al., 2013). We suggest that users of NLP applications should be empowered to make informed decisions that are consistent with their moral worldviews. NLP projects should aim to be transparent about dependencies on NLP resources that have been developed with religious considerations in mind, in line with calls for greater transparency around datasets and models generally (e.g., Bender and Friedman, 2018; Gebu et al., 2021; Mitchell et al., 2019).

Indigenous cultures. Language, religion and culture are not orthogonal or separable. Given the

NLP community’s skew towards the West (Rungta et al., 2022), the values of, and possible harms to, local communities of diverse cultures are not known by most NLP researchers. For Indigenous NLP projects using translations of sacred texts, we echo calls for more consideration of local colonial contexts, to consider community opinions, and for research to prioritize the needs of Indigenous communities (Bird, 2020; Schwartz, 2022; Alvarado Garcia et al., 2021). Mager et al. (2023) demonstrate one way in which community opinions can be sought regarding NLP projects, and deeper relationships with communities will provide more insights. NLP researchers working with Indigenous languages should become familiar with Indigenous perspectives summarized in Section 4, and with codes of conduct such as the ones published by the Endangered Languages Project.¹¹ Janke (2021) provides useful guidelines concerning Indigenous Cultural and Intellectual Property (ICIP). With the imminent extinction of many Indigenous languages and religions, we also suggest there may be a role for NLP to play in documenting and maintaining “first-person accounts of what people once believed in and how they talked to and about their gods” (Harrison, 2007, p. 153).

6 Conclusion

This position paper presented a study of the use of religious texts in NLP research, finding that common scenarios are dataset creation and machine translation. We have argued that responsible secularism requires the ACL to engage with concerns about how such NLP activities might impact religious communities, especially the most powerless ones, or might be complicit with projects which do. We provided a detailed account of some of the considerations, with a focus on Indigenous cultures and communities, and suggested how the field can more responsibly engage with questions of religious positionality and cultural standpoints.

7 Researcher Positionality

I live and work in a secular, English-speaking, colonized country of the Global North. I grappled while writing this paper with my lack of first-hand experiential understanding of religion, and thus too with my personal role in arguing for more consideration of global and local religious communities.

¹⁰As of September 2023, the ACL Anthology has no mentions of “language positionality” nor “linguistic positionality”.

¹¹https://fpcc.ca/wp-content/uploads/2023/02/CodeOfConduct_Web.pdf

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