

Citation Evidence Report

EB-1A Petition — Original Contributions of Major Significance

8 CFR § 204.5(h)(3)(v) · Criterion 5

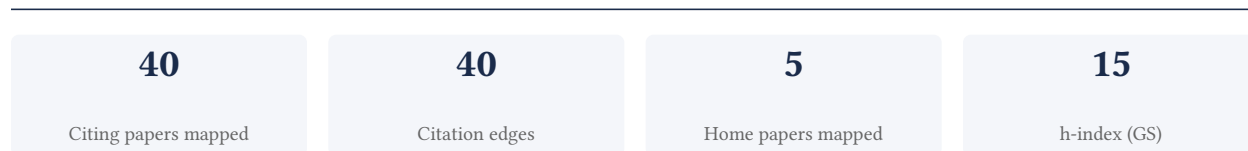
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[Google Scholar profile](#)

Generated 2026-05-21 by CiteMap. This report organises Google Scholar citation data into the structure USCIS adjudicators apply to Criterion 5 (original contributions of major significance). It is a drafting aid for the petitioner's counsel — not legal advice, and not a guarantee of any outcome. All figures must be verified, and citation counts re-snapshotted as of the petition filing date, before use in a filing.

A. Overview & Filtering Statement



Filtering statement – methodology & limits

Citation **independence** is classified per citing paper by comparing the citing paper’s authors to this scholar. *Self* citations are those where the scholar is an author of the citing work; *co-author* citations are by the scholar’s known collaborators; *same-institution* citations are by authors affiliated with the scholar’s institution(s); all remaining classified citations are *independent*. Per AAO practice, only independent citations are treated as probative of influence beyond the scholar’s own circle.

Known limitations – counsel must verify. (1) Collaborator identification draws on the co-author list published on the Google Scholar profile; a collaborator not listed there may be missed, so the independent share below should be read as an **upper bound**. (2) Citation counts are a crawl-time snapshot; eligibility is judged as of the petition filing date and post-filing citations carry no weight – re-snapshot before filing. (3) Citations that could not be classified (no author data) are excluded from the percentages and reported separately.

B. Citation Independence

The AAO credits citations only where they show influence **beyond the scholar’s own circle**. Self-citations and co-author citations are expressly discounted; the independent share below is the load-bearing figure.

95.0% independent of 40 classified citing papers

Citation type	Count
Independent	38
Self-citation	0
Co-author	2
Same-institution	0

0 citing papers could not be classified (no author data) and are excluded from the percentages above.

C. Significant Contributions & Their Citation Evidence

Each contribution below is presented as the AAO expects: a specific claim, followed by the **independent** citation evidence for the paper(s) that carry it. Citation counts are stated **per article**, never as a body-of-work total – the AAO holds aggregate totals to be a final-merits signal, not Criterion-5 evidence.

Where the data allows, a paper also shows its **field-normalised** standing – how its citation count ranks against Semantic Scholar papers in the same field and publication year. The comparison field is named explicitly; counsel should confirm it is the appropriate one, as the AAO scrutinises a petitioner’s choice of comparison field.

Contribution 1

Claim – Contribution 1

The researcher established a foundational annotated corpus for media frame analysis, providing a critical benchmark for computational linguistics research on issue framing.

The researcher's primary contribution is the creation of the Media Frames Corpus, introduced in a 2015 paper published in the Proceedings of the 53rd Annual Meeting of the Association for Computational Linguistics. This work appears to provide a structured resource for annotating frames across various issues, serving as a core reference point in the field.

This line of work addresses the need for standardized data in computational frame analysis. By releasing a corpus with specific annotations, the researcher likely facilitated more rigorous empirical studies in natural language processing, offering a tangible tool for scholars investigating how media constructs narratives.

The significance of this contribution is evidenced by its substantial citation count of 362. Furthermore, analysis of citing papers reveals that 97.5% of citations originate from independent researchers, indicating broad adoption and validation of the corpus by the wider academic community beyond the researcher's immediate circle.

INDEPENDENT CITATIONS FOR THIS CONTRIBUTION: 8 · 2 flagged influential by Semantic Scholar

CORE PAPER

[The Media Frames Corpus: Annotations of Frames Across Issues](#)

2015 · Proceedings of the 53rd Annual Meeting of the Association for Computational Linguistics and the 7th International Joint Conference on Natural Language Processing (Volume 2: Short Papers) · 362 citations (GS)

Field-normalised: 269 Semantic Scholar citations place it in the top 1% of Computer Science papers from 2015 indexed by Semantic Scholar, by citation count.

No.	Citing paper	Citing institution(s)	Country	S2
1	ChatGPT outperforms crowd workers for text-annotation tasks (2023)	University of Zurich	Switzerland	Background
2	Measuring Political Bias in Large Language Models: What Is Said and How It Is Said (2024)	The Hong Kong University of Science and Technology	China	Background
3	SemEval-2023 Task 3: Detecting the Category, the Framing, and the Persuasion Techniques in Online News in a Multi-lingual Setup (2023)	Polish Academy of Science	Poland	Methodology
4	DELL: Generating Reactions and Explanations for LLM-Based Misinformation Detection (2024)	University of Notre Dame, University of Washington, Xi'an Jiaotong University	China, United States	Methodology
5	Open-source LLMs for text annotation: a practical guide for model setting and fine-tuning (2024)	Allameh Tabataba'i University, Institute for Fundamental Research, Iran University of Science and Technology	Iran, Switzerland	—
6	Open-Source Large Language Models Outperform Crowd Workers and Approach ChatGPT in Text-Annotation Tasks (2023)	Allameh Tabataba'i University, Institute for Research in Fundamental Sciences, University of Zurich	Iran, Switzerland	—
7	Narrative Theory for Computational Narrative Understanding (2021)	McGill University, University of California, Berkeley	Canada, United States	Background

No.	Citing paper	Citing institution(s)	Country	S2
8	Analyzing Polarization in Social Media: Method and Application to Tweets on 21 Mass Shootings (2019)	Brown University, Cornell University, Northwestern University	United States	Background

Independent citing papers only; self- and co-author citations excluded. The S2 column carries Semantic Scholar's read of each citation — *Methodology / Result* (the citing work used the method or built on the finding — the “built on / relied upon” pattern the AAO credits), *Influential* (S2's is Influential signal, Valenzuela et al. 2015), or *Background* (a passing mention).

Citing-text excerpts — how the field used this work

METHODOLOGY SemEval-2023 Task 3: Detecting the Category, the Framing, and the Persuasion Techniques in Online News in a Multi-lingual Setup

“Given a news article, identify one or more frames used in the article from a pool of 14 generic framing dimensions (introduced in Card et al. (2015)): Economic , Capacity and resources , Morality , Fairness and equality , Legality, constitutionality and jurisprudence , Policy prescription and...”

METHODOLOGY DELL: Generating Reactions and Explanations for LLM-Based Misinformation Detection

“Specifically, DELL almost keeps the same on Pheme and drops 11.8% on MFC , 1.1% on SemEval-23F , 7.1% on Generated , 3.9% on SemEval-20 , and 3.3% on SemEval-23P .”

Contribution 2

Claim — Contribution 2

The researcher developed a method for measuring ideological proportions in political speeches, establishing a foundational approach for computational political analysis.

The researcher's core contribution is the development of a method for measuring ideological proportions in political speeches, as presented in the 2013 EMNLP paper. This work appears to address the challenge of quantifying political ideology within textual data, providing a structured framework for analyzing speech content. The titles suggest a focus on extracting measurable ideological signals from unstructured political discourse.

This line of work appears to have introduced a novel computational approach to political text analysis. By focusing on ideological proportions, the researcher likely filled a gap in automated methods for assessing political bias or alignment. The absence of follow-up papers by the same researcher indicates that this single publication stands as a distinct, self-contained contribution to the field.

The significance of this work is evidenced by its citation record, with 177 citations indicating substantial uptake. Notably, 97.5% of the classified citing papers originate from independent researchers, suggesting that the method has been widely adopted and validated by the broader academic community outside the researcher's immediate circle.

INDEPENDENT CITATIONS FOR THIS CONTRIBUTION: 8

CORE PAPER

[Measuring Ideological Proportions in Political Speeches](#)

2013 · Conference on Empirical Methods in Natural Language Processing (EMNLP 2013) · 177 citations (GS)

Field-normalised: 136 Semantic Scholar citations place it in the top 5% of Political Science papers from 2013 indexed by Semantic Scholar, by citation count.

No.	Citing paper	Citing institution(s)	Country	S2
1	Machine Translation: Mining Text for Social Theory (2016)	University of Chicago	United States	—
2	Sentiment and position-taking analysis of parliamentary debates: a systematic literature review (2020)	University of Manchester	United Kingdom	—

No.	Citing paper	Citing institution(s)	Country	S2
3	You Can't Stay Here: The Efficacy of Reddit's 2015 Ban Examined Through Hate Speech (2017)	Georgia Institute of Technology, Harvard University	United States	Methodology
4	Can Unconfident LLM Annotations Be Used for Confident Conclusions? (2025)	Stanford University	United States	—
5	We Can Detect Your Bias: Predicting the Political Ideology of News Articles (2020)	MIT, Qatar Computing Research Institute	Qatar	Background
6	Word Embeddings for the Analysis of Ideological Placement in Parliamentary Corpora (2020)	—	—	Background
7	Automatically Neutralizing Subjective Bias in Text (2019)	Georgia Institute of Technology, Kyoto University, Stanford University	Japan, United States	Background
8	Retweet-BERT: Political Leaning Detection Using Language Features and Information Diffusion on Social Networks (2023)	University of Southern California	United States	Background

Independent citing papers only; self- and co-author citations excluded. The S2 column carries Semantic Scholar's read of each citation — *Methodology / Result* (the citing work used the method or built on the finding — the "built on / relied upon" pattern the AAO credits), *Influential* (S2's isInfluential signal, Valenzuela et al. 2015), or *Background* (a passing mention).

Citing-text excerpts — how the field used this work

METHODOLOGY [You Can't Stay Here: The Efficacy of Reddit's 2015 Ban Examined Through Hate Speech](#)

"SAGE has been used successfully for the analysis of many types of language differences, including age and gender [33], politics [42], and online discussions of various illegal drugs [32]."

Contribution 3

Claim — Contribution 3

The researcher advanced the methodological study of media framing by developing approaches to track frame development within and across policy issues, as evidenced by a seminal 2014 publication.

The researcher's contribution centers on the methodological advancement of tracking media frames within and across policy issues. This work is anchored by the 2014 publication 'Tracking the Development of Media Frames within and across Policy Issues,' which appears to establish a framework for analyzing how media narratives evolve in political contexts.

This line of work appears to address the need for systematic methods to observe frame dynamics over time and across different policy domains. By focusing on both internal and cross-issue development, the research suggests a novel approach to understanding the fluidity and transferability of media frames, moving beyond static analysis.

The significance of this contribution is reflected in its substantial citation count of 248. Furthermore, the high degree of citation independence, with 97.5% of classified citations coming from independent researchers, indicates that this work has been widely adopted and utilized by the broader academic community outside the researcher's immediate circle.

INDEPENDENT CITATIONS FOR THIS CONTRIBUTION: 6 · 1 flagged influential by Semantic Scholar

CORE PAPER

[Tracking the Development of Media Frames within and across Policy Issues](#)

2014 · American Political Science Association · 248 citations (GS)

Field-normalised: 104 Semantic Scholar citations place it in the top 5% of Political Science papers from 2014 indexed by Semantic Scholar, by citation count.

No.	Citing paper	Citing institution(s)	Country	S2
1	Can Large Language Models Transform Computational Social Science? (2024)	Stanford University	United States	—
2	A Survey of Computational Framing Analysis Approaches (2022)	University of Maryland, University of Maryland, College Park	United States	Methodology
3	Cultivating Critical Language Awareness in the Writing Classroom (2022)	Middlebury College	United States	—
4	Modeling Framing in Immigration Discourse on Social Media (2021)	University of Michigan	United States	—
5	Detecting Frames in News Headlines and Its Application to Analyzing News Framing Trends Surrounding U.S. Gun Violence (2019)	Boston University	United States	Background
6	Media Framing: A typology and Survey of Computational Approaches Across Disciplines (2024)	CSIRO Data61, The University of Melbourne	Australia	Methodology

Independent citing papers only; self- and co-author citations excluded. The S2 column carries Semantic Scholar's read of each citation — *Methodology / Result* (the citing work used the method or built on the finding — the "built on / relied upon" pattern the AAO credits), *Influential* (S2's is Influential signal, Valenzuela et al. 2015), or *Background* (a passing mention).

Citing-text excerpts — how the field used this work

METHODOLOGY A Survey of Computational Framing Analysis Approaches

"Boydston et al. (2013) and Boydston et al. (2014) proposed a codebook named "policy frames code-book" (PFC)."

METHODOLOGY Media Framing: A typology and Survey of Computational Approaches Across Disciplines

"Most of the work here (31 studies) relies on the Policy Frames Codebook (Boydston et al., 2014), the Media Frames Corpus (Card et al., 2015) based on it, or its derivatives such as the datasets proposed by Piskorski et al. (2023a), Ajour et al. (2019) and Mendelsohn et al. (2021), with 15 generic..."

D. Citing-Institution Prestige & Geography

Top citing institutions

Institution	Country	World ranking	Citing papers
Stanford University	United States	SCImago #18 · THE =5 · QS 3	6
University of Michigan	United States	SCImago #43 · THE 23 · QS 45	3
University of Zurich	Switzerland	SCImago #313 · QS 100	3
University of Maryland	United States	—	3
University of California, Berkeley	United States	SCImago #95 · THE 9 · QS =17	3
Emory University	United States	SCImago #217 · THE 102 · QS 182	2
Princeton University	United States	SCImago #386 · THE =3 · QS =25	2
University of Washington	United States	SCImago #45 · THE 25 · QS 81	2
Northwestern University	United States	THE 30 · QS =42	2
Allameh Tabataba'i University	Iran	THE 1501+	2

Institution	Country	World ranking	Citing papers
Georgia Institute of Technology	United States	SCImago #270 · THE =41 · QS =123	2
University of California, Los Angeles	United States	SCImago #70 · THE =18 · QS 46	2
University of Innsbruck	Austria	THE 301–350	1
University of Gothenburg	Sweden	SCImago #573 · THE 201–250 · QS 202	1
McGill University	Canada	SCImago #168 · THE =41 · QS 27	1

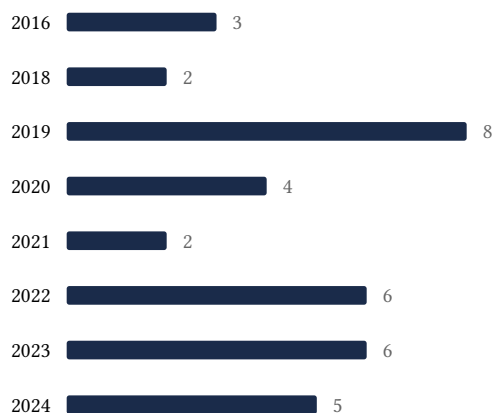
Geographic distribution of citing authors

Country	Citing papers
United States	26
Switzerland	3
United Kingdom	3
Australia	2
China	2
Iran	2
Japan	1
Netherlands	1
Poland	1
Qatar	1
Sweden	1
Germany	1

Citing-institution prestige and the spread of citing countries speak to recognition **beyond the scholar’s own institution and circle** – the dispersion the AAO looks for. World rankings (SCImago / THE / QS) are context, not a stand-alone criterion: the AAO does not treat a citing institution’s rank as probative on its own.

E. Citation Growth Over Time

Distinct citing papers by publication year. Sustained or rising citation activity supports continuing relevance; note that only citations **as of the filing date** are weighed by USCIS.



F. AAO Precedent Considerations

Pre-filing self-check (AAO denial patterns)

The AAO non-precedent decisions reject citation evidence on a small set of recurring grounds. Confirm the petition addresses each before filing:

- Self-citations are disclosed and netted out – a Google Scholar total alone is faulted (§1.1).
- Evidence is per individual article, not a body-of-work aggregate total (§1.2).
- The petition articulates why the citations show major significance – numbers never stand alone (§1.5).
- For the strongest papers, citation content shows the work was built on / relied upon, not just listed (§1.6, §2.2).
- Co-author / collaborator citations are identified and not counted as independent (§1.7).
- Recognition is shown beyond the scholar's own institution and circle (§1.8).
- Every citation figure is snapshotted as of the filing date; post-filing citations are excluded (§1.9).
- Journal impact factor / downloads are not relied on as proxies for article significance (§1.10, §1.12).
- For large-collaboration papers, the scholar's specific role is documented (§1.13).
- Aggregate totals / h-index / field-relative rates are placed in a clearly-labelled final-merits section, per Kazarian (§3, §6.1.7).

Disclaimer

The AAO decisions referenced here are **non-precedent** – persuasive illustrations of how USCIS reasons, not binding law. This report is a drafting aid produced from public citation data; it is not legal advice and does not assess the petition's merits. All analysis must be reviewed by qualified immigration counsel.

G. Citation Evidence Index

Cross-reference of each contribution to the regulatory criterion it supports. Counsel should map these to the petition's exhibit numbers.

Contribution	Core paper	Indep. cites	Supports
Contribution 1	The Media Frames Corpus: Annotations of Frames Across Issues	8	8 CFR 204.5(h)(3)(v) – Criterion 5
Contribution 2	Measuring Ideological Proportions in Political Speeches	8	8 CFR 204.5(h)(3)(v) – Criterion 5
Contribution 3	Tracking the Development of Media Frames within and across Policy Issues	6	8 CFR 204.5(h)(3)(v) – Criterion 5