

Academic Affiliations amongst Philosophy Departments

Walter Barta

University of Houston

As Presented at the 2019 Digital Research Commons Workshop

Data Collected November 2019 [Updated November 2023]

ABSTRACT

The prestige of an academic institution may be determined as a function of affiliations with other academic institutions. Using digital tools to data-scrape, data-mine, and perform network analysis on university websites, an approximation of numbers of academic affiliations may be measured. Especially observing the alma mater institutions of the faculty of employed institutions, these numbers show the relative employment of alumni and a proxy metric for the relative prestige of their degree-granting institutions. These affiliations can be charted and graphed to determine the distributions of affiliations throughout an academic ecosystem from which we might draw conclusions about that system's hierarchies and inequalities. Here we use anglophone PhD-granting philosophy departments as a case study for this methodology with tentative conclusions.

INTRODUCTION

Different academic institutions have different amounts of institutional prestige. University of Houston has more institutional prestige than Houston Community College. But just how is prestige distributed across academia?

Many have attempted to assess and analyze the relative prestige of academic institutions. Some news organizations like USnews and Forbes have yearly studies based on a variety of factors combined to produce a concise listing of the academies ("National University Ranking"; "America's Top College List"). Some organizations including the MLA and other independent research groups keep running reports on the job placements of college graduates by program (MLA; Colander et al., 2015).

These ranking systems incorporate different factors as proxies of institutional prestige. For instance, the USnews website lists over ten factors going into their academic ranking ("National University Ranking"). But something that USNews and others leave out is the metric of "academic affiliation", the degree of connection between institutions in the academic network, more specifically the frequency with which PhD graduates of academic institutions are hired as professors by other academic institutions. Arguably, academic affiliation, via the intermural hiring process, is the primary process by which academic prestige is distributed, since it implicitly involves the granting and receiving of prestige between academic institutions in the form of human capital (Menand, 2010). This is as much as to say, how academic an academy may be is determined by how academic other academies consider it. Doctors beget doctors, engineers beget engineers, professors beget professors. In the academic sense, academies are only as good as other academies say they are. Thus, in theory, the prestige of an academic

institution should be roughly determinable through the proxy of frequency of academic affiliation with other academic institutions.

In what follows here, we attempt to use digital tools of data-scraping, data-mining, and network analysis on university websites to make an approximation of numbers of academic affiliations amongst anglophone PhD-granting philosophy departments, use those metrics to infer institutional prestige, and use the distribution of institutional prestige to make some claims about the academic community.

METHODOLOGY

Summary

The number of outgoing and incoming affiliations of a university was determined by taking a survey of university websites. “Affiliations” were the designated unit of measurement, defined as an instance of occurrence of the name of an academic institution on the website of another academic institution. (These represented a cross-reference between institutions just as a URL link represents a cross-reference between websites.) Thus, each affiliation was comprised of two components: 1) the “source”, the outgoing affiliation, the academic institution from which the affiliation points; and 2) the “target”, the incoming affiliation, the academic institution to which the affiliation points. The amount of “targets” of any given institution shows the relative attention that institution receives amidst other institutions.

Procedure

Breaking the process down into detailed methodology, the following steps were followed:

1. A list of anglophone PhD-granting philosophy departments and their website URLs was compiled from “APA Grad Guide Search Results”.
2. A list of departmental faculty members (full and associate professors) and their directory sub-URLs was compiled from each departmental website.
3. A list of alma mater institutions was compiled from each faculty directory website sub-URL.
4. A query was run to produce a “Target/Source” list, where each “Target” was a institution at which a given faculty member works and each “Source” the a institution from which that faculty member received their doctorate
5. A query was run to produce a “Count/Rank” list, where each “Count” was the raw cardinal number of times a degree-granting institution in the data was cited and where “Rank” was the ordinal ranking of the count from greatest to least.
6. The “Target/Source” text file was input into a Gephi network graph and an “Academic Affiliations Network” file was produced.
7. The “Count/Rank” text file was input into an Excel bar graph and an “Academic Affiliations Distribution” file was produced.

Scope

These steps were conducted across a scope of targets and sources. The targets were determined from a range of 123 anglophone PhD-granting Philosophy Programs selected from the APA Grad Guide list (“APA Grad Guide Search Results”). This scope was chosen for three main reasons:

- There is a small (thus achievable) number of PhD-granting philosophy programs.
- Philosophy as a discipline is internally interested in critically analyzing the hierarchy/inequality of social distributions.
- Philosophy is the first academic discipline and therefore a natural first focus of our study (which can and may expand outward to encompass more disciplines given continued resources and interest).

Tools

The digital tools used were Python Selenium for data-scraping, Microsoft Access for data-mining, and Gephi and Microsoft Excel for Network Analysis.

Errors

Many errors (both random and systematic) affected the fidelity of the results. These errors included, but were surely not limited to:

- Missing affiliations: Some online academic profiles lacked affiliation information. This is estimated to be as much as 10% of all profiles, but we can reasonably assume the missing information was random (thus representative) and therefore not significantly skewing of results.
- Multiple affiliations: Some online academic profiles had more than one academic affiliation. This was a vanishingly small number and therefore vanishingly skewing..
- Ambiguous labels: Some academic affiliations had ambiguous labels corresponding to more than one academic institution. These cases were mostly resolved using educated guesses, and might skew results relating to ambiguously named institutions, but not others.
- Mis-processing: The data-collection, data-cleaning, and data-analyzing steps may have introduced manual or automated errors into the data set.

RESULTS

The steps above produced results reflective of the academic affiliations of institutional websites (The results are shown in the following tables in Appendix A: Tables and figures in Appendix B: Figures).

- “Table 1: Academic Affiliation Sources/Targets (Selected)” lists the academic affiliations of institutions in terms of target institution and source institution. Each instance in the table represents an instance of a target institution documented from a source institution source website, each source/target relation representing one “academic affiliation”.
- “Table 2: Academic Affiliation Counts/Ranks (Top Twenty-Five)” lists institutions in order of academic affiliation count and rank. Using the “Targets/Sources” table, the number of affiliations of which an institution was a target was counted to produce a “target count”.
- “Table 3: Academic Affiliation Ranks (Selected)” lists institutions in order of academic affiliation rank. Using the “Target Count” table, the number of affiliations of which an institution was a target was arranged in order to produce an algorithmic rank. These

results were then compared to the Forbes rank from the Forbes website to produce a rank difference, by which the results might be compared against the Forbes study.

- “Figure 1: Academic Affiliation Networks” shows the network of institutions according to their academic affiliations. The nodes represent institutions and the edges represent academic affiliations. Those institutions with smaller numbers of affiliations were smaller nodes; those institutions with larger numbers of affiliations were larger nodes.
- “Figure 2: Academic Affiliation Distribution” shows the pareto distribution of institutions according to their academic affiliations. The x-axis represents different academic institutions and the y-axis represents numbers of “target affiliations” of those given institutions. Those institutions with higher numbers of affiliations had taller bars in the chart while those institutions with lower numbers of affiliations had shorter bars in the chart. The resulting curve indicates academic inequality: a linear curve would indicate a very egalitarian academic community, a steeply indented curve would indicate an unequal academic community.

DISCUSSION

At least two interesting conclusions can be drawn from these results:

Academic Affiliations are Proxies for Institutional Prestige

The metric of “academic affiliation” is a valid proxy of institutional prestige in as much as it is consistent with other metrics of prestige. Those institutions with large/small numbers of affiliation “targets” have large amounts of academic prestige. This can be deduced from the results in two ways:

Firstly, the results are consistent with our common knowledge expectations about the relative prestige of universities. Princeton and other Ivy League universities have a strong affiliated lead whereas on average state schools tend to be middling to less affiliated.

Secondly, and more importantly, the rankings of institutions in one study can be compared to the rankings produced by another study, for example those offered by academic rankings organizations like forbes.com, etc. (“America’s Top College List”). The results of academic affiliation are shown to be relatively consistent with the results of at least one other study such that academic affiliation is a valid proxy of institutional prestige at least in the minimal sense that it can reproduce the results of other studies of institutional prestige. Although the exact relative ranking varies, the margin of error is relatively low over all (Table 3).

Academic Affiliation Distributions are Unequal (not Unexpectedly)

The distribution of prestige amongst academic institutions is steeply unequal. The academic affiliation distribution has a high Gini inequality coefficient: $\sim 0.7/1.0$ (Gini). This can be compared with the distribution of wealth in the American economy with has a relatively lower Gini inequality coefficient: $\sim 0.5/1.0$ (census.gov). This is the pattern that one would expect for the peak of a social distribution (Pareto, 1898), to the extent that philosophy PhDs are a demographic at the peak of performance in their field. This further suggests that, like other unequal market systems, academia is not immune to differentially distributing socioeconomic forces (Menand, 2010). This all can be observed visually from the stark shape of the rankings curve (Figure 2).

CONCLUSION

These results are interesting for their implications for how to think about prestige economies in general and those of academic institutions in particular. How the academy thinks about its own processes of student admissions, faculty hiring, and university funding might be reassessed according to the concept of academic affiliation and its association with prestige. Further studies of this kind discovering the network patterns amongst academic institutions could be carried out both to extend the breadth (the types of academic programs: history, biology, etc.) and depth (the levels of academic programs: associate programs, bachelor's programs, master's programs, etc.) of the scope, such that academic affiliations could be a more thoroughly and comprehensively studied and applied subject.

ACKNOWLEDGEMENTS

Special thanks to David Mazella, Claude Willan, Elizabeth Stravoski, Graham Lee and other members of the community at the M.D. Anderson Library Digital Research Commons at University of Houston for feedback on the project.

APPENDIX A: TABLES*Table 1: Selected Academic Affiliation Sources/Targets (Selected)*

Target	Source
Arizona State University	Arizona State University
Arizona State University	Arizona State University
Arizona State University	Arizona State University
Arizona State University	Arizona State University
Arizona State University	Arizona State University
Arizona State University	City University of New York (CUNY) Graduate Center
Arizona State University	Claremont Graduate School
Arizona State University	Harvard University
Arizona State University	New York University
Arizona State University	Rutgers University
Arizona State University	Stanford University
Arizona State University	University of Arizona
Arizona State University	University of Arizona
Arizona State University	University of British Columbia
Arizona State University	University of California, Irvine
Arizona State University	University of California, Santa Barbara
Arizona State University	University of Maryland-College Park
Arizona State University	University of Maryland-College Park
Arizona State University	University of Notre Dame
Arizona State University	University of Southern California
Arizona State University	University of Texas at Austin
Arizona State University	N/A
Arizona State University	N/A
Arizona State University	N/A
Arizona State University	N/A

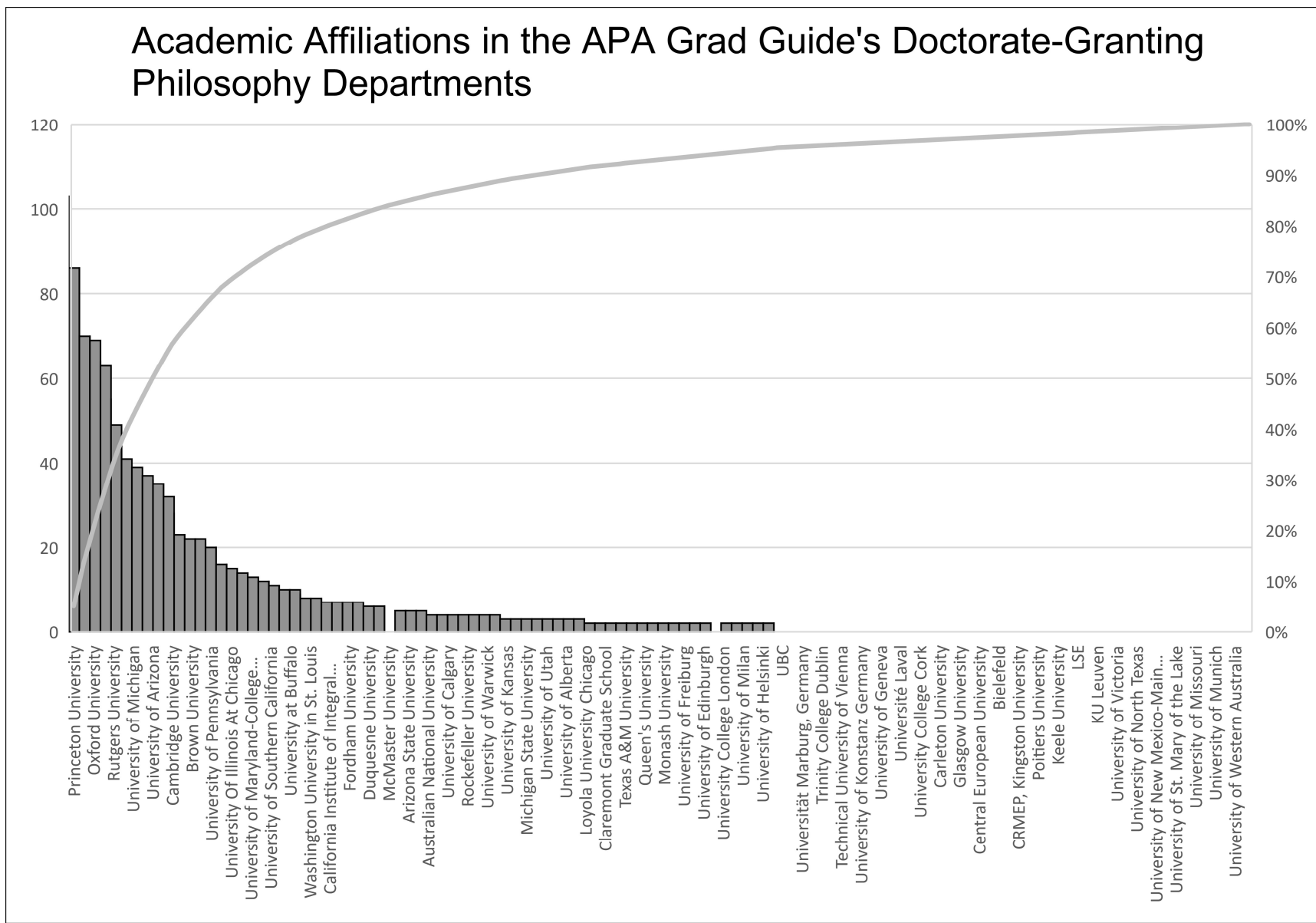
Table 2: Academic Affiliation Target Counts (Top Twenty-Five)

Affiliation	Count	Rank
Princeton University	103	1
University of Pittsburgh - Philosophy	86	2
Harvard University	86	3
University of Toronto	70	4
Oxford University	70	5
The University of Chicago	69	6
University of California, Berkeley	64	7
Massachusetts Institute of Technology	64	8
Rutgers University	63	9
Yale University	55	10
University of California-Los Angeles	49	11
University of Notre Dame	44	12
University of Michigan	41	13
Columbia University	39	14
Cornell University	39	15
New York University	37	16
University of Arizona	37	17
University of Texas at Austin	36	18
Stanford University	35	19
University of North Carolina Chapel Hill	35	20
Cambridge University	32	21
University of California-San Diego	28	22
City University of New York (CUNY) Graduate Center	24	23
University of Wisconsin-Madison	23	24
Brown University	23	25

Table 3: Academic Affiliation Rank (Selected)

Name	Rank	ForbesRank	RankDifference
Princeton University	1	5	0.615384615
Harvard University	3	1	-0.307692308
University of Pittsburgh - Philosophy	3	151	22.76923077
The University of Chicago	6	18	1.846153846
Massachusetts Institute of Technology	8	4	-0.615384615
University of California, Berkeley	8	14	0.923076923
Rutgers University	9	136	19.53846154
Yale University	10	2	-1.230769231
University of California-Los Angeles	11	46	5.384615385
University of Notre Dame	12	21	1.384615385
University of Michigan	13	22	1.384615385
Cornell University	15	13	-0.307692308
Columbia University	15	15	0
University of Arizona	17	200	28.15384615
New York University	17	48	4.769230769
University of Texas at Austin	18	74	8.615384615
University of North Carolina Chapel Hill	20	47	4.153846154
Stanford University	20	3	-2.615384615
University of California-San Diego	22	81	9.076923077
City University of New York (CUNY) Graduate Center	23	178	23.84615385
University of Wisconsin-Madison	25	75	7.692307692
Brown University	25	8	-2.615384615
Penn State University	28	121	14.30769231
Stony Brook University	28	182	23.69230769
University of California, Irvine	28	96	10.46153846
University of Pennsylvania	29	7	-3.384615385
Northwestern University	31	20	-1.692307692
Indiana University	31	132	15.53846154
Boston University	33	78	6.923076923
Emory University	34	52	2.769230769

Figure 2: Academic Affiliation Distribution



REFERENCES

- “2022 Income Inequality Decreased for First Time Since 2007.”
<https://www.census.gov/library/stories/2023/09/income-inequality.html>
- “America’s Top College List.” <https://www.forbes.com/top-colleges/list/>
- “APA Grad Guide Search Results.” <https://gradguide.apaonline.org/search-results-export?eid=148035&token=-it2a0EbnjJnELS2i2PBTEghA-F-QX9yGO6jxSNw5i4&return-url=program-search>
- Brin, Sergey; Page, Lawrence. “The Anatomy of a Large-Scale Hypertextual Web Search Engine.” Computer Science Department, Stanford University, Stanford, CA.
- Colander, David; Zhou, Daisy. “Where Do PhDs in English Get Jobs?: An Economist’s View of the English PhD Market.” *Pedagogy*, Volume 15, Issue 1, January 2015, pp. 139-156. Duke University Press.
- Gini, C. (1909). "Concentration and dependency ratios" (in Italian). English translation in *Rivista di Politica Economica*, **87** (1997), 769–789.
- Grafton, Anthony T., and Jim Grossman. October 2011, “No More Plan B: A Very Modest Proposal for Graduate Programs in History.” *Perspectives on History*, October. www.historians.org/perspectives/issues/2011/1110/1110pre1.cfm .
- “How U.S. News Calculated the 2019 Best Colleges Rankings.”
<https://www.usnews.com/education/best-colleges/articles/how-us-news-calculated-the-rankings> .
- Menand, Louis. 2010. *The Marketplace of Ideas*. New York: Norton.
- MLA Office of Research. September 2011. “Report on the MLA ‘Job Information List,’ 2010 – 11,” Modern Language Association, www.mla.org/pdf/sept_rpt_jil_1011_v2.pdf.
- “National University Ranking.” <https://www.usnews.com/best-colleges/rankings/national-universities>
- Pareto, Vilfredo (1898). "Cours d'economie politique". *Journal of Political Economy*. **6**. doi:10.1086/250536.
- Smith, Sidonie. 2010. “Beyond the Dissertation.” *MLA Newsletter*, Spring. www.mla.org/blog&topic=133 .