



University of Kentucky
UKnowledge

Markey Cancer Center Faculty Publications

Cancer

3-15-2018

Evaluating Disparities in the U.S. Technology Transfer Ecosystem to Improve Bench to Business Translation

James Weis

Massachusetts Institute of Technology

Ashvin Bashyam

Massachusetts Institute of Technology

Gregory J. Ekchian

Massachusetts Institute of Technology

Kathryn Paisner

KP2 LLC

Nathan L. Vanderford

University of Kentucky, nathan.vanderford@uky.edu

Right click to open a feedback form in a new tab to let us know how this document benefits you.

Follow this and additional works at: https://uknowledge.uky.edu/markey_facpub

 Part of the [Policy Design, Analysis, and Evaluation Commons](#), [Science and Technology Studies Commons](#), and the [Translational Medical Research Commons](#)

Repository Citation

Weis, James; Bashyam, Ashvin; Ekchian, Gregory J.; Paisner, Kathryn; and Vanderford, Nathan L., "Evaluating Disparities in the U.S. Technology Transfer Ecosystem to Improve Bench to Business Translation" (2018). *Markey Cancer Center Faculty Publications*. 113. https://uknowledge.uky.edu/markey_facpub/113

This Article is brought to you for free and open access by the Cancer at UKnowledge. It has been accepted for inclusion in Markey Cancer Center Faculty Publications by an authorized administrator of UKnowledge. For more information, please contact UKnowledge@lsv.uky.edu.

Evaluating Disparities in the U.S. Technology Transfer Ecosystem to Improve Bench to Business Translation

Notes/Citation Information

Published in *F1000Research*, v. 7, 329, p. 1-18.

© 2018 Weis J *et al.*

This is an open access article distributed under the terms of the [Creative Commons Attribution Licence](#), which permits unrestricted use, distribution, and reproduction in any medium, provided the original work is properly cited.

Digital Object Identifier (DOI)

<https://doi.org/10.12688/f1000research.14210.1>



RESEARCH ARTICLE

Evaluating disparities in the U.S. technology transfer ecosystem to improve bench to business translation [version 1; referees: 3 approved, 1 approved with reservations]

James Weis ^{1-4*}, Ashvin Bashyam^{1,2,5*}, Gregory J. Ekchian^{1,5,6*}, Kathryn Paisner⁷, Nathan L. Vanderford ^{8,9}

¹MIT Biotechnology Group, Massachusetts Institute of Technology, Cambridge, MA, USA

²Department of Electrical Engineering & Computer Science, Massachusetts Institute of Technology, Cambridge, MA, USA

³Computational & Systems Biology Initiative, Massachusetts Institute of Technology, Cambridge, MA, USA

⁴Computer Science & Artificial Intelligence Laboratory, Massachusetts Institute of Technology, Cambridge, MA, USA

⁵Koch Institute for Integrative Cancer Research, Massachusetts Institute of Technology, Cambridge, MA, USA

⁶Department of Materials Science & Engineering, Massachusetts Institute of Technology, Cambridge, MA, USA

⁷KP2 LLC, Oakland, CA, USA

⁸Markey Cancer Center, University of Kentucky, Lexington, KY, USA

⁹Department of Toxicology and Cancer Biology, College of Medicine, University of Kentucky, Lexington, KY, USA

* Equal contributors

v1 First published: 15 Mar 2018, 7:329 (doi: [10.12688/f1000research.14210.1](https://doi.org/10.12688/f1000research.14210.1))
 Latest published: 15 Mar 2018, 7:329 (doi: [10.12688/f1000research.14210.1](https://doi.org/10.12688/f1000research.14210.1))

Abstract

Background: A large number of highly impactful technologies originated from academic research, and the transfer of inventions from academic institutions to private industry is a major driver of economic growth, and a catalyst for further discovery. However, there are significant inefficiencies in academic technology transfer. In this work, we conducted a data-driven assessment of translational activity across United States (U.S.) institutions to better understand how effective universities are in facilitating the transfer of new technologies into the marketplace. From this analysis, we provide recommendations to guide technology transfer policy making at both the university and national level.

Methods: Using data from the Association of University Technology Managers U.S. Licensing Activity Survey, we defined a commercialization pipeline that reflects the typical path intellectual property takes; from initial research funding to startup formation and gross income. We use this pipeline to quantify the performance of academic institutions at each step of the process, as well as overall, and identify the top performing institutions via mean reciprocal rank. The corresponding distributions were visualized and disparities quantified using the Gini coefficient.

Results: We found significant discrepancies in commercialization activity between institutions; a small number of institutions contribute to the vast majority of total commercialization activity. By examining select top performing institutions, we suggest improvements universities and technology transfer offices could implement to emulate the environment at these high-performing institutions.

Conclusion: Significant disparities in technology transfer performance exist in which a select set of institutions produce a majority share of the total

Open Peer Review

Referee Status:

	Invited Referees			
	1	2	3	4
version 1				
published 15 Mar 2018	report	report	report	report

- Karen I. Deak**, University of Notre Dame, USA
- Adriana Bankston** , Future of Research, USA
- Evan Facher**, University of Pittsburgh, USA
- Dipanjan Nag**, The Ohio State University, USA
Christopher Paschall, The Ohio State University, USA

Discuss this article

technology transfer activity. This disparity points to missed commercialization opportunities, and thus, further investigation into the distribution of technology transfer effectiveness across institutions and studies of policy changes that would improve the effectiveness of the commercialization pipeline is warranted.

Comments (0)

Keywords

Commercialization, Technology Transfer, Technology Licensing, Patents, Licenses, Startups

Corresponding author: Nathan L. Vanderford (nathan.vanderford@uky.edu)

Author roles: **Weis J:** Conceptualization, Data Curation, Formal Analysis, Investigation, Methodology, Project Administration, Supervision, Validation, Visualization, Writing – Original Draft Preparation, Writing – Review & Editing; **Bashyam A:** Conceptualization, Formal Analysis, Investigation, Methodology, Software, Validation, Visualization, Writing – Original Draft Preparation, Writing – Review & Editing; **Ekchian GJ:** Conceptualization, Formal Analysis, Investigation, Methodology, Validation, Visualization, Writing – Original Draft Preparation, Writing – Review & Editing; **Paisner K:** Validation, Writing – Review & Editing; **Vanderford NL:** Conceptualization, Investigation, Project Administration, Resources, Supervision, Writing – Review & Editing

Competing interests: No competing interests were disclosed.

How to cite this article: Weis J, Bashyam A, Ekchian GJ *et al.* **Evaluating disparities in the U.S. technology transfer ecosystem to improve bench to business translation [version 1; referees: 3 approved, 1 approved with reservations]** *F1000Research* 2018, 7:329 (doi: [10.12688/f1000research.14210.1](https://doi.org/10.12688/f1000research.14210.1))

Copyright: © 2018 Weis J *et al.* This is an open access article distributed under the terms of the [Creative Commons Attribution Licence](https://creativecommons.org/licenses/by/4.0/), which permits unrestricted use, distribution, and reproduction in any medium, provided the original work is properly cited.

Grant information: The author(s) declared that no grants were involved in supporting this work.

First published: 15 Mar 2018, 7:329 (doi: [10.12688/f1000research.14210.1](https://doi.org/10.12688/f1000research.14210.1))

Introduction

The transfer of inventions from academic institutions to private industry is a major driver of economic growth and human welfare. Broadcom, Google, Akamai, Yahoo, Biogen, Bose, and Genentech represent just a handful of pioneering companies with academic roots (Kenney, 2017). Indeed, many of today’s defining technologies originated in academic labs, including nuclear energy and the internet (Busbin, 1995; Manyika & Roxburgh, 2011; Nelson & Byers, 2015).

Technology-driven progress demands not only the development of new inventions, but also their dissemination throughout society. Our national capacity to fuel growth and improve human well-being through new technologies depends on our ability to pass these technologies through a commercialization pipeline. This national need for an efficient and effective technology handoff between academia and industry motivated our analysis of the current United States (U.S.) academic technology transfer environment.

Leveraging data from the Association of University Technology Managers (AUTM) U.S. Licensing Activity Survey, we characterized the performance of research organizations across different steps of the technology transfer process. Our findings indicate that the translational abilities of research organizations across the U.S. vary widely, with a small minority of institutions producing the vast majority of technological and economic benefits. To begin addressing this gap, we surveyed initiatives aimed at improving technology transfer and propose remedies for observed disparities in institutional performance.

Methods

Defining the commercialization pipeline

The AUTM Licensing Survey solicits responses annually from around 300 institutions, including universities, hospitals and research institutions, to quantify the total technology transfer activity at these institutions. These metrics are derived from a set of core questions that AUTM deems essential for assessing transfer and licensing activity. A detailed description of each metric from the AUTM survey data is given in Supplementary Table 1. We defined the “commercialization pipeline” (Figure 1) by identifying a set of key questions asked in each AUTM survey, and extracting relevant data from the 2010 to 2014 AUTM surveys. We use this commercialization pipeline to measure and compare relative levels of technology transfer activity at different institutions, and at different steps along the pipeline. The distributions of each metric across every surveyed institution are visualized as linear and log histograms, as well as empirical cumulative distributions, in Supplementary Figure 1 and Supplementary Figure 2.

Identification of top performing institutions

We ranked each institution from the AUTM Licensing Survey data by each step in the commercialization pipeline. Any institution ranked in the top 10 (about the top 5%) in at least one stage of the pipeline was included in our list of top performing institutions. This resulting list of 25 institutions (approximately 12% of all surveyed institutions) was then sorted based on mean reciprocal rank (MRR):

$$MRR = \frac{1}{N} \sum_i \frac{1}{R_i}$$

where $N = 7$ is the number of stages in the pipeline and R_i is the ranking of the institution in step i of the pipeline. We chose this scoring system to identify institutions with consistently high performance across the commercialization pipeline while avoiding heavily penalizing anomalous weak performances in just a single metric.

Calculation of the Gini coefficient

Given value x_i for institution i and x_j for institution j , we calculate G , the Gini Coefficient, such that:

$$G = \frac{\sum_{i=1}^n \sum_{j=1}^n |x_i - x_j|}{2 \sum_{i=1}^n \sum_{j=1}^n x_j} = \frac{\sum_{i=1}^n \sum_{j=1}^n |x_i - x_j|}{2n \sum_{i=1}^n x_i}$$

The Gini coefficient is a measure of statistical dispersion used to assess inequality in a population. A high Gini coefficient indicates high levels of inequality where, in this case, a few institutions contribute a substantial amount of total translational activity. Conversely, a low Gini coefficient indicates that each institution contributes an equal share.

Statistical analysis

Variance estimates (v) for the Gini coefficient for each step were derived via jackknife resampling (Karagiannis & Kovacevic’, 2000; Yitzhaki, 1991):

$$v = \frac{N-1}{N} \sum_{i=1}^N (G_i - G)^2$$

where N is the number of observations, G is the Gini coefficient when all observations are considered, and G_i is the Gini coefficient value when the i th observation is removed. The confidence intervals of the log-normal fits were computed to the 95% confidence levels using the Jacobian of the parameter



Figure 1. Commercialization pipeline. Each step in this pipeline corresponds to a metric in the AUTM survey. We use the health of the pipeline as a proxy for the overall health of the U.S. technology transfer ecosystem.

estimates assuming normally distributed residuals. All statistical analysis was performed in **MATLAB** 2016b (Mathworks, Natick, MA, USA).

Results

Inequality between institutions through pipeline

Our goal was to understand how much each institution contributed to each step of the commercialization pipeline and to determine any notable overall trends in U.S. technology transfer. Histograms (**Supplementary Figure 1** and **Supplementary Figure 2**) of contributions from each institution along the commercialization pipeline reveal highly skewed distributions. The distributions of each metric are generally well approximated by a log-normal fit. Note that the x-axis is on a log scale and therefore the significant skew in the distribution is not immediately apparent. The effectiveness of a log-normal fit decreases towards the end of the commercialization pipeline (Startups and Adjusted Gross Income).

The majority of institutions contribute a small amount to overall technology transfer regardless of how activity is measured (**Figure 2** and **Supplementary Figure 3**). Specifically, the top 20% of institutions contribute over 60% of total commercialization activity. Importantly, this trend is robust to normalization by research expenditures, which indicate that differences in research funding do not explain the gap in productivity (**Supplementary Figure 3**). In fact, the top 10% of institutions

contribute over 40% of “startups per dollar of research expenditures” and over 70% of “adjusted gross income per dollar of research expenditures”.

Highly performing institutions

We identified the 25 top-performing institutions by sorting all top-performing institutions by the average of their reciprocal ranking at each step in the commercialization pipeline (**Table 1**). Most organizations that perform well do so across the entire commercialization pipeline, indicating strong and broad technology transfer abilities (e.g. University of California and University of Texas Systems; MIT; and Stanford). On the other hand, some organizations excel in only specific parts of the commercialization pipeline (e.g. University of Washington in Licenses and Options Executed; California Institute of Technology in New Patent Applications; and University of Georgia in Licenses and Options Executed), which reveals focused, less-robust technology transfer capabilities.

Dispersion analysis

We extended this analysis by calculating the Gini coefficient, a measure of statistical dispersion that is often used to quantify income inequality (**Gini, 1912**). In this analysis, a low Gini coefficient indicates that each institution is contributing roughly equally to U.S. technology commercialization, whereas a high Gini coefficient indicates that a few institutions are producing the majority of the commercialization output.

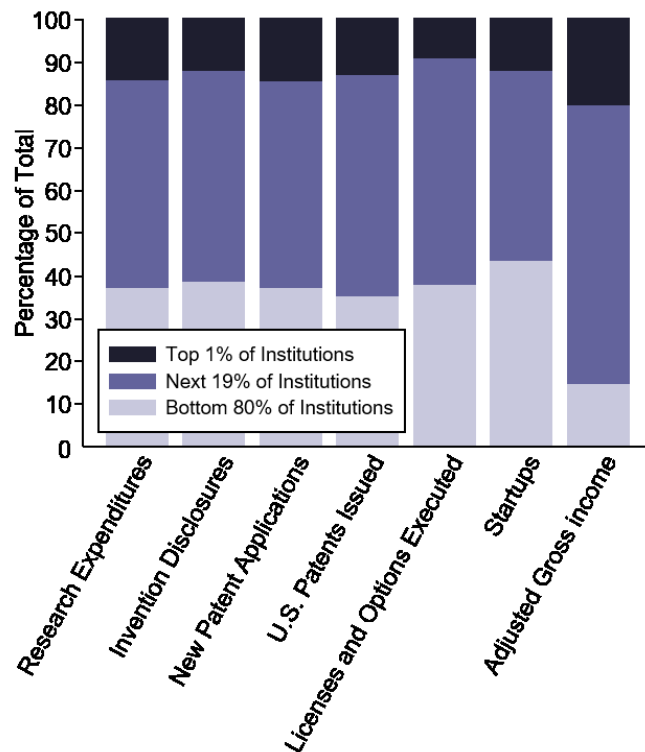


Figure 2. Contribution by top 1%, top 20% and bottom 80% of institution to each step of the commercialization pipeline. A small number of institutions contribute to the majority of commercialization activity.

As shown in [Figure 3](#) and [Supplementary Figure 3](#), high levels of inequality exist throughout the pipeline. For context, the Gini coefficient of patents issued in the U.S. is above 60%, while the Gini coefficient of all U.S. household income is 48% ([U.S. Census Bureau](#)). We believe this indicates that the majority of U.S. research organizations have significant untapped commercialization potential, the full realization of which could lead to new technologies and, overall, improved U.S. productivity.

Improving the pipeline

Many of the top performing institutions have invested significant effort and resources in supporting entrepreneurs at each stage of the commercialization pipeline. Top performing institutions have ensured continuity in their support structure to enable the efficient and effective translation and development of both institute-owned and student-created intellectual property. [Table 2](#) highlights active programs at MIT and Harvard, two top performing translational institutions. Our summary of these initiatives span university incubators, student organizations, university venture capital funds and business plan competitions ([Table 2](#)).

The overview of successful programs ([Table 2](#)) provides a blueprint for universities that would like to foster improved technology transfer and innovation. While some of these programs would require a significant undertaking on the part of the university, many can be achieved in a straightforward and lightweight manner via the support of student-led activities and partnership with government and private organizations. Examples of grassroot student groups that have launched many new programs exist at both MIT and Harvard. For instance, the MIT Biotech Group group has partnered with the MIT Alumni Angels of Boston to launch a life sciences-focused track to improve access to capital for early-stage startups. The Harvard Biotechnology Club runs an incubator program to develop and translate academic research. These programs represent student-led efforts that require little to no university expenditure or resources. For larger undertakings, university/corporate collaborations can provide an efficient means to achieve significant progress. A prime example of this is JLABS @ M2D2, the medical device incubator partnership between Johnson & Johnson and the University of Massachusetts Lowell ([McCarthy et al., 2013](#)).

Table 1. The 25 top-performing institutions. Bar plots show the mean value over the years under consideration for each institution for each step in our commercialization pipeline.

Institution	Research Expenditures (\$M)	Invention Disclosures	New Patent Applications (US)	Patents Issued (US)	Licenses and Options Executed	Startups
University of California System	5364	1605	1117	359	261	66
University of Texas System	2508	772	357	173	155	24
Massachusetts Inst. of Technology (MIT)	1515	646	514	232	107	18
Stanford University	855	492	308	210	116	16
Johns Hopkins University	1540	417	400	72	132	10
University of Washington/Wash. Res. Foundation	1010	401	172	75	225	12
California Institute of Technology	426	389	565	149	51	10
University of Michigan	1257	363	152	106	115	11
UW-Madison/WARF	1113	378	129	153	63	6
University of Pennsylvania	886	386	196	78	107	15
Columbia University	737	356	230	86	82	15
University of Illinois, Chicago, Urbana	972	354	159	102	87	12
Massachusetts General Hospital	744	326	185	83	132	9
University of Florida	548	330	167	84	128	14
Cornell University	781	363	175	82	130	10
University of Utah	401	223	99	67	82	18
University of Georgia	307	163	56	34	157	3
Georgia Institute of Technology	741	364	228	79	72	10
Harvard University	812	377	213	66	78	9
University of Colorado	809	239	297	39	53	9
University System of Maryland	998	292	180	64	38	9
Duke University	845	212	127	47	118	6
University of Pittsburgh	755	264	89	49	124	6
University of South Florida	441	177	89	89	58	9
Johns Hopkins University Applied Physics Laboratory	1101	219	59	18	29	3

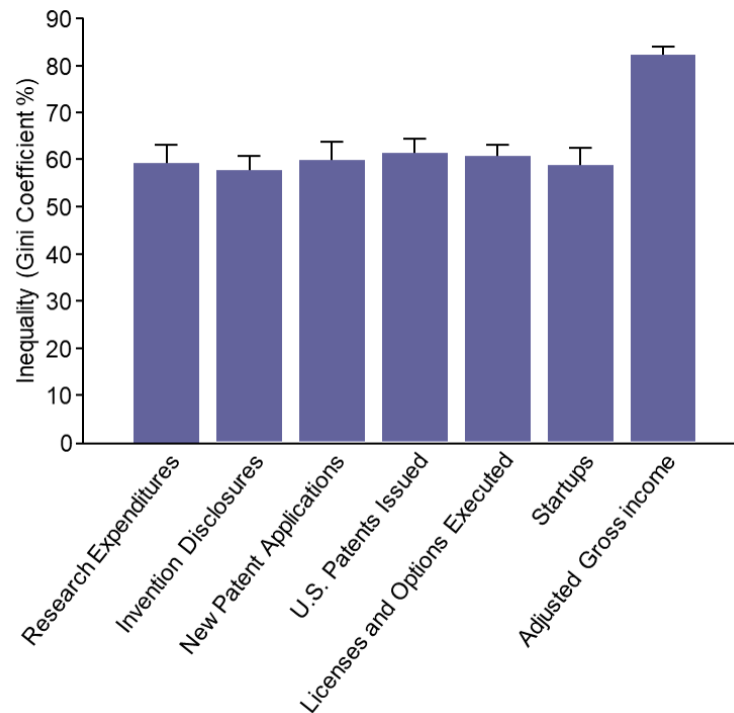


Figure 3. The Gini Coefficient for each stage in the commercialization pipeline, with G of 0% representing complete equality and G of 100% represents complete inequality. Error bars represent one standard deviation of uncertainty as estimated via jackknife resampling (Karagiannis & Kovacevic, 2000; Yitzhaki, 1991).

Discussion

Expense, time, infrastructure, and the lack of partnerships are among the most common barriers to research commercialization and alleviating these bottlenecks allows more inventions to enter the marketplace (Vanderford *et al.*, 2013). Programs to increase support for inventors at less well performing institutions to file disclosures, pursue patent prosecution, and seek licensing deals could significantly boost translational output. Sharing best practices from the leaders in technology commercialization may help bring more new technologies to market.

Supporting the commercialization pipeline

One salient feature of the top-performing institutions is their broad portfolio of commercialization-focused initiatives. Individually, these projects typically target only a few steps on our commercialization pipeline (for example, business plan competitions target the latter stages of the technology transfer process). However, the best performing universities have a large number of these efforts which, in aggregate, fully span the commercialization pipeline. This observation indicates a potential strategy for improvement of those less well served technology transfer pipelines; specifically, the cultivation of commercialization focused initiatives, such as incubators, business plan competitions, innovation prizes, law clinics, and student organizations. The value of these efforts goes beyond their immediate impact. For example, although when taken at face value, a business plan

competition may seem to serve only the winning team, its merit truly stems from bringing together students, entrepreneurs, investors, and the media in a constructive setting. The resources required for such projects are small, and, given the disparity in commercialization, potential societal benefits are vast.

Outsourcing technology transfer to a third-party

A clear barrier to effective commercialization of university technology is the widespread lack of access to experienced, motivated, and well-resourced technology transfer offices (TTO). Many institutions are unable to support a comprehensive TTO, hampering efforts to introduce new technology into industry. The use of consultants can help alleviate some shortcomings, but faces its own barriers to widespread adoption (AUTM Technology Transfer Practice Manual).

Alternatively, a coalition of institutions could create a third-party technology licensing organization whose charter is to serve the technology transfer needs of those institutions. Like a sports agent, this third-party organization would use its expertise to strike technology transfer deals between institutions and licensees, freeing universities to focus on their strengths. Funded directly by the institutions and, in part, by licensing revenue, this organization would have the necessary resources and freedom to hire top-tier technology transfer professionals who can effectively interface between stakeholders in industry and in academia,

Table 2. Current programs at MIT and Harvard, two of the top-performing institutions, that strengthen the commercialization pipeline. The shaded regions denote which areas of the pipeline each program most directly addresses.

	Research Expenditures	Invention Disclosures	New Patent Apps (U.S.)	Patents Issued (U.S.)	Licenses and Option Executed	Startups	Adjusted Gross Income
Harvard			Harvard Life Labs A 15,000-square-foot shared laboratory space for high-potential life sciences and biotech startups founded by Harvard				
			President's Innovation Challenge A Campus-wide competition supporting student ventures through networking events, mentorship and funding				
		Harvard Catalyst Program An NIH funded center fostering a translational environment enabling collaboration and providing tools, training and technologies to investigators					
			Harvard Biotechnology Club A student organization that hosts events and provides educational services that allow members to explore the world of business and biotechnology				
MIT			The Engine An institute-backed venture capital fund empowering disruptive technologies with the long-term capital, knowledge, and specialized equipment and labs they need to thrive				
			\$100K Entrepreneurship Competition A student run entrepreneurship contest offering mentorship from venture capitalists, serial entrepreneurs, corporate executives, and attorneys, media exposure, prototyping funds, business plan feedback, and discounted services				
			Sloan Healthcare Innovation Prize A student run pitch competition supporting early-stage healthcare startups with feedback from industry professionals, pitch workshops, and funding				
		Sandbox An institute-backed seed funding for student-initiated entrepreneurship ideas, mentoring from both within and outside of MIT, and tailored educational experiences					
			Deshpande Center for Technological Innovation Support for bringing early-stage technologies to the marketplace in the form of breakthrough products and new companies through grants, mentorship, industry connections, and an annual symposium				
			MIT Biotechnology Group A student organization that buildings strong, symbiotic relationships between the MIT community, academia, and industry and serves the MIT community by facilitating development of knowledge, skills, networks, and experiences to prepare members for biotechnology-related careers				
					MIT Life Sciences Alumni Angels An alumni run angel investing network that supports MIT startups focused on the life sciences with funding, connections, and mentorship		

while negotiating on behalf of the parent institutions. These teams would work to creatively package and license technologies to maximize their utility to society, as well as to assure that the parent institutions receive a fair return on their investment.

Operating outside of the university, this organization would be free to make decisions much more quickly than traditional TTOs. Similarly, its employees would be incentivized to work in the best interest of the parent institutions by ensuring the process is both

efficient and maximizes value for all stakeholders. This outsourced model of technology transfer speaks towards the latent need for more efficient, properly incentivized, and more widespread efforts to commercialize academic research and development efforts.

Conclusion

As the U.S. economy becomes increasingly driven by technological change, understanding and improving the commercialization pipeline is critically important. The significant disparity in technology transfer performance is evident as the top few institutions produce a very large share of the country's total technology transfer. We believe this disparity points to missed commercialization opportunities, which we as a society are paying for by missing out on potentially highly impactful innovations.

Data availability

The AUTM Licensing Activity Survey data are available on the organization's website (<https://www.autm.net/resources-surveys/research-reports-databases/licensing-surveys/>) by fee or institutional subscription/membership. As such, the raw data analyzed for this study cannot be provided in the context of this article. The 2010–2014 survey data used for this study was obtained as part of an institutional membership (University of Kentucky).

Competing interests

No competing interests were disclosed.

Grant information

The author(s) declared that no grants were involved in supporting this work.

Supplementary material

Supplementary Figure 1. Histograms of each step in the commercialization pipeline shown in [Figure 1](#). Insets show cumulative distributions, with shaded rectangles indicating the number of institutions necessary to reach 80% of total activity.

[Click here to access the data.](#)

Supplementary Figure 2. Histograms of each step in the commercialization pipeline shown in [Figure 1](#) with a log x-axis. Note that the x-axis is on a log scale and therefore the significant skew in the distribution is not immediately apparent.

[Click here to access the data.](#)

Supplementary Figure 3. Quantized and normalized distribution for each step in the commercialization pipeline. The shaded bars represent the percentage of the total of each category owned by institutions in percentiles indicated.

[Click here to access the data.](#)

Supplementary Table 1. Pipeline descriptions.

[Click here to access the data.](#)

References

- Association of University Technology Managers **Technology Transfer Practice Manual**. 3rd ed. [Reference Source](#)
- Busbin JW: **The Role of Marketing Research and Decision Systems in the Marketing Process**. *Journal of Nonprofit & Public Sector Marketing*. 1995; 2(2–3): 167–90. [Publisher Full Text](#)
- Gini C: **Variabilità E Mutabilità, Contributo Allo Studio Delle Distribuzioni E Delle Relazioni Statistiche: Fascicolo Ier: Introduzione - Indici Di Variabilità - Indici Di Mutabilità**. 1912. [Reference Source](#)
- Karagiannis E, Kovacevic' M: **A Method to Calculate the Jackknife Variance Estimator For the Gini Coefficient**. *Oxford Bulletin of Economics and Statistics*. 2000; 62(1): 119–22. [Publisher Full Text](#)
- Kenney M: **The Diverse Roles of Universities in Regional Innovation Ecosystems: Case Studies from University of California Campuses***. 2017.
- McCarthy SP, Tello S, Hafer N: **Medical Device Innovation in Massachusetts-M2D2: Massachusetts Medical Device Development Center**. UMass Center for

Clinical and Translational Science Seminar Series. October 17 2013.

[Reference Source](#)

Manyika J, Roxburgh C: **The Great Transformer: The Impact of the Internet on Economic Growth and Prosperity**. McKinsey Global Institute, October 2011.

[Reference Source](#)

Nelson A, Byers T: **Organizational Modularity and Intra-University Relationships between Entrepreneurship Education and Technology Transfer**. In *Advances in the Study of Entrepreneurship, Innovation & Economic Growth*. 2015; 275–311.

[Publisher Full Text](#)

U.S. Census Bureau: **Current Population Survey, 1968 to 2015 Annual Social and Economic Supplements**.

[Reference Source](#)

Vanderford NL, Weiss LT, Weiss HL: **A survey of the barriers associated with academic-based cancer research commercialization**. *PLoS One*. 2013; 8(8): e72268. [PubMed Abstract](#) | [Publisher Full Text](#) | [Free Full Text](#)

Yitzhaki S: **Calculating Jackknife Variance Estimators for Parameters of the Gini Method**. *Journal of Business & Economic Statistics: A Publication of the American Statistical Association*. 1991; 9(2): 235. [Publisher Full Text](#)

Open Peer Review

Current Referee Status:



Version 1

Referee Report 11 April 2018

doi:10.5256/f1000research.15458.r32326



Dipanjan Nag , Christopher Paschall

The Ohio State University, Columbus, OH, USA

Referee #1

- The authors correctly point out that this is a skewed industry, with the top performing schools in any given category contributing more than their share of returns. (Page 1)
- The selection of top 10 institutions in the initial analysis and then expanding to top 25 institutions is somewhat arbitrary. (Page 3)
- This data is coming from institutions of all sizes that vary according to the research dollars as well as the overall size of institutions. For an objective analysis the data needed to be normalized along those lines. (Page 3)
- Some of the stages described by the authors in Figure 1 are metrics that can be manipulated. For example the number of patents applied for and issued really are not metrics of measuring efficiency or impact of an office. Large number of patents can be applied for and granted, the real measure of performance will be the number of patents that are licensed. (Page 3)
- The approach of creating a composite metric is a good one, rather than just looking at the revenue number. The Milken study is another one that has utilized this approach and is an objective measure. The difference is that the Milken study normalized the data, as well as provided the weighting of each factor. (Page 3)
- In most cases the large revenue numbers can be attributed to a single large license or a monetization activity from a university. The weighting of that single license will skew the result enormously. Combined with the lack of data normalization, the results will not be reflective of the true impact to technology commercialization from universities. The weighting factor for each of the "steps" and how they overall impact the ranking is unclear. The weighting of the input metrics vs. the output metrics cannot be the same, hence the rankings would change quite radically. That explanation should be very clearly pointed out. (Page 4)
- The majority of institutions here are based on their size and hence a normalization is needed to avoid picking institutions by size only. Smaller schools often times are more efficient.
- In looking at Table 1 their analysis shows the 1% and the top 19% percent of the institutions. There are 300 institutions in the survey. Why did they pick 25 to rank?
In addition Table 1 shows that the top 20% account for 60% of results. Why didn't they cite the top 60 institutions. It would seem logical to look at the top 20%. (Table 1)
- The sharing of best practices has existed for decades in the very collaborative technology transfer community. The AUTM Leadership Conference is precisely that effort, along with the LES IP100 meeting. The fact is that, it is not easy to solicit inventions from faculty. It takes years if not decades to get them on board with the idea that commercialization is a good thing. (Page 6)

- These programs are ideal in nurturing an "ecosystem" for technology commercialization. The authors are absolutely right about taking a holistic approach rather than just trying to increase disclosures, patents, startups or revenues. (Page 6)
- The whole idea of hiring consultants to fix technology transfer does not work. I have personally been a consultant and a technology transfer professional for many years. The notion of fixing tech transfer through consultant is not economically viable. Consultants most of the time do not have the necessary understanding of the university structure and functions. Many aspects of technology transfer will not have economic impact or metrics based impact, but they need to be done nonetheless. Analogy of hiring consultant would be like replacing full time and dedicated servicemen with mercenaries. Having said that, hiring of consultants for focused engagements is highly recommended and successful offices do that often times in areas such as patent prosecution, valuation, licensing comparable, or market analysis. If there are significant examples that the authors can cite to support the replacement of technology transfer offices with consultants, that would cause readers to consider such an approach. Short of that authors should either reconsider adding this to the article or reframing the recommendation. (Page 6)
- This model has been tried in Japan, where both TTOs and TLOs existed. The model failed miserably. Not only is this model not economically viable, but it creates an atmosphere of trying to create a consortium out of disparate companies like Amazon, Google, Ford and IBM. Some consortiums on a limited scope such as a particular technology focus, e.g. IoT, with a few institutions could work. Again, the authors should cite specific examples that clearly demonstrate the validity of the proposed model otherwise reconsider adding this as a recommendation. (Page 6)
- The picking of two institutions which are ranked according to the authors at #3 and #19 does seem highly biased analysis. Additionally, the programs that the authors have described are student and faculty programs. Technology transfer performed by universities across the board and the data reported by AUTM do not take into account any student programs. Hence, the impact that authors have contemplated at each stage are possibly not accurate. Suggest the inclusion of universities from different "ecosystems" as well as exclusion of student programs from the analysis. (Page 7)
- From a practical standpoint this model will be at best difficult if not impossible with the interest of the parties diverging because of a simple reason, royalty stacking. Each inventor or assignee wants to have the highest value for their portion of the invention. That is just one factor from an economic standpoint. There are many other intangible factors that would be barriers. Anyone who has setup consortia can attest to these challenges. (Page 7)
- Overall, the article has significant challenges:
 1. Authors are predominantly from the Boston area and are analyzing their own institution and another institution which is in close proximity. There are more than 300 institutions and a number of thriving "ecosystems". This is a significant bias and in some ways looks more like a marketing piece as opposed to an objective article. The reason the two institutions are successful could be because of the existence of industry players, local VCs, selective admission of students at two private colleges, and many other factors.
 2. Their analysis seems to have consisted solely of sorting the AUTM data, not much analysis (interviews, or other supporting data) about the drivers of those outcomes. Hence, they don't show any support for their statements about technology transfer offices being under performing.
 3. The recommendations or solutions that the authors are proposing are not supported by any data or examples and are simply anecdotal. If such models have existed and worked then analysis of those models would be helpful.

4. As we all know the Milken study was published with different results than the authors present. They do not reference the Milken study or explain the differences. Other articles which have attempted to perform similar analysis have cited the Milken study. (e.g. Mature Biotech Entrepreneur in 2014)

5. Technology transfer is indeed a complex interplay of a various factors that are not always easy to understand. The intangible factors such as location, concentration of research areas, public versus private institution, can all have impact that cause significant disparity in the results. Authors need to consider analyzing or at least mentioning some of those factors in this article. (Page 8 - Conclusion)

Referee #2

1. The years 2010-2014 is a small, five-year sample of data, from which broader trends and conclusions are difficult to draw. The industry moves and changes rapidly, so the 2014 data (which is FY2014, or July 2013 – June 2014 for most institutions), is a relatively old data set.
2. Why were only the top 10 in each category of the Commercialization pipeline included in the MRR analysis?
3. The parameters identified in the Commercialization pipeline (Figure 1) to measure the health of a tech transfer office are, in many instances, inadequate or irrelevant. If one is trying to measure the health of a *university* (not simply its TTO), then one might use research expenditures as a reasonable variable. However, this reviewer fails to see the direct correlation between an institution's research expenditures and health of the TTO. Furthermore, technology commercialization is a non-linear, dynamic process. The notion that commercialization proceeds from left-to-right (i.e. research expenditure-invention-patent-license-startup-revenue) is incomplete.
4. Similarly, with respect to patent applications and patents issued, such numbers depend heavily on an individual institution's IP strategy. For example, California Institute of Technology, files provisional patent applications on nearly every invention disclosure that is received. How does such a policy speak to the health of the TTO?
5. Licenses and options are certainly one reasonable metric that can be used to assess the health of a TTO. However, all licenses and options are not created equal. Some universities will license 30 technologies or patents in a single license, and some will license the same 30 technologies or patents in 30 different licenses. Is the second university healthier than the first? The authors have failed to address or account for differences in quality of licenses and options between universities. The quantities by themselves don't paint an accurate picture.
6. For the Startups metric, the authors should address, again, the quality of the startups coming out of the universities. Some universities create large numbers of startups in a given year, but some of those may not be legitimate, growth-oriented startups with qualified management.
7. Adjusted Gross Income should not be used as a metric to measure the health of TTOs. The TTO has no control over this value.
8. In the Results section on page 4, I find it disappointing that the authors go through so much analysis to demonstrate the truth of the Pareto principle in technology transfer. The fact that 20% of the universities contribute to 60% of "total commercialization activity" is entirely unsurprising.
9. Again, in the Results section on page 4, "Highly performing institutions", I don't see that the data were normalized for research expenditures. If so, the analysis needs to be rerun. Furthermore, the authors need to remove the University of California, University of Texas, and University of Maryland systems because those systems report to AUTM statewide. Alternatively, the authors could retrieve any desired data from the individual institutions in those states.

10. I am not clear on the analogy between the Gini coefficient of patents and US household income. How do the authors conclude that a 60% Gini coefficient for patents means that universities have significant untapped commercialization potential?
11. Page 5, "Improving the pipeline", along with Table 2, reads more like an advertisement for Harvard and MIT, particularly the Biotechnology Clubs of which some of the authors are members. The paper would be strengthened if they identified activities conducted outside of Boston-based universities that are contributing unexpected positive results to the tech transfer ecosystem.

Discussion section comments

1. Sharing best practices across institutions is a common and frequent activity in the university tech transfer community. Organizations like AUTM exist for such a reason.
2. Where do the authors propose universities obtain resources to promote additional commercialization activity?
3. Do the authors have any data to support one of their proposed solutions, the outsourcing of technology transfer to consultants?

Overall questions and comments

1. Specifically, outside of Harvard and MIT, what are institutions in the top 25 doing that other institutions can replicate to improve their TTOs?
2. Are the authors aware of similar programs at other universities as those highlighted in Table 2?
3. Have the authors considered non-traditional factors that may contribute to the success of TTOs, such as overall university mission, geographic location, appetite for risk, patience for technology development timelines, etc.?

References

1. Weis J, Bashyam A, Ekchian G, Paisner K, Vanderford N: Evaluating disparities in the U.S. technology transfer ecosystem to improve bench to business translation. *F1000Research*. 2018; 7. [Publisher Full Text](#)

Is the work clearly and accurately presented and does it cite the current literature?

Partly

Is the study design appropriate and is the work technically sound?

Partly

Are sufficient details of methods and analysis provided to allow replication by others?

Partly

If applicable, is the statistical analysis and its interpretation appropriate?

I cannot comment. A qualified statistician is required.

Are all the source data underlying the results available to ensure full reproducibility?

Yes

Are the conclusions drawn adequately supported by the results?

No

Competing Interests: No competing interests were disclosed.

We have read this submission. We believe that we have an appropriate level of expertise to confirm that it is of an acceptable scientific standard, however we have significant reservations, as outlined above.

Referee Report 09 April 2018

doi:10.5256/f1000research.15458.r32093



Evan Facher

Innovation Institute, University of Pittsburgh, Pittsburgh, PA, USA

The authors tackle a very relevant question virtually all technology commercialization organizations at academic institutes have – what can they do to enhance translation of innovations developed at their organizations. In order to assess the current situation and provide some thoughts on approaches to ameliorate the challenges their colleague institutes have, the authors took a well reasoned, data intensive approach to examine a large set of information. The AUTM data is a valuable source of self-reported figures and a large amount of it was assessed to reach the presented conclusions. The design was appropriate and relevant to answer the hypotheses posed in this article. Further, a fairly robust statistical analysis was performed. Given the data is available (for purchased access) to any institute of higher learning, the analyses described in the paper can be readily replicated, and potentially expanded on, by any group that seeks to build on it.

The conclusions drawn as a result of this assessment are supported by the data and provide multiple areas for practioners of technology transfer to examine further for implementation. I believe the manner in which the authors broke down the commercialization pipeline was sound and is a good model for others to use in assessing their own processes. The only suggestion I would offer is that universities are increasingly focusing on software licensing, as a result, not all translation will be of patented innovations but rather copyrights. As such, there may be additional information that can be garnered for these inventions. One additional area that may be interesting to include for the authors if they chose to eventually expand on the analysis beyond this submission is understanding the disparity of human resources at the AUTM schools to understand how much, if any, this specific infrastructure component is tied to commercialization activity and AUTM metrics.

In conclusion, I enjoyed the opportunity to review this article and believe it adds to the literature on technology transfer. A significant amount of data was used to identify recommendations that align with opportunities most organizations can adopt.

Is the work clearly and accurately presented and does it cite the current literature?

Yes

Is the study design appropriate and is the work technically sound?

Yes

Are sufficient details of methods and analysis provided to allow replication by others?

Yes

If applicable, is the statistical analysis and its interpretation appropriate?

Yes

Are all the source data underlying the results available to ensure full reproducibility?

Yes

Are the conclusions drawn adequately supported by the results?

Yes

Competing Interests: No competing interests were disclosed.

I have read this submission. I believe that I have an appropriate level of expertise to confirm that it is of an acceptable scientific standard.

Referee Report 06 April 2018

doi:[10.5256/f1000research.15458.r32556](https://doi.org/10.5256/f1000research.15458.r32556)



Adriana Bankston 

Future of Research, Abington, MA, USA

General comments:

This publication highlights an important topic in examining the effectiveness of universities to facilitate the transfer of new technologies into the marketplace. This work also raises awareness to the fact that only a small number of institutions contribute to the vast majority of the total commercialization activity. This is only one of the many disparities found across U.S. institutions, and the work presented here very effectively puts this particular disparity into a broader context.

Therefore, this type of research has the potential to be utilized as a way to press for change in the system as a whole. Particularly for this field, a change in the system would require advocating for the implementation of successful practices from high-performing institutions across a higher number of U.S. institutions, so that eventually these technologies can become commonplace in all universities. To this end, I believe it is particularly important to consider how we might provide similar types of resources to U.S. institutions with less of an access to these tools or with fewer number of tools at their disposal as compared to the high-performing institutions in the area of technology transfer.

The publication itself is very well-written, with the motivation for pursuing this work and its purpose being very clearly defined. The authors have also very succinctly articulated the gap filled by this work and its overall significance to the field, and the results support the conclusions drawn. Overall, I believe this paper is suitable for publication in its current form. I have just a few suggestions for improvement, or ideas that might be useful to consider if expanding or revising this paper in the future. These are mostly broad recommendations related to whether and how the disparities described here could be reduced across universities.

General considerations:

I would like to commend the authors for detailing the steps of the commercialization pipeline early in the publication - this is very helpful for those who might not be very familiar with it. Similarly, the explanations of the data analysis parameters, including the Gini coefficient, are very useful in understanding the conclusions drawn from these studies. One suggestion I would have is to move the information from the Supplementary Table 1 into a main table in the publication, to enable the reader to more easily and quickly refer to these concepts in conjunction with reading Figure 1.

I also appreciate the fact that the commercialization pipeline allows the authors to answer key questions using relevant data in the AUTM Licensing Survey. However, it would be really helpful to also conduct a more in-depth investigation of how each of the universities in the AUTM Licensing Survey measures up in every step of the commercialization pipeline, in order to obtain a view of the national landscape on this issue beyond just the top 25 institutions. While I realize that looking at all 300 universities may be cumbersome and that analyzing the top 25 was a relevant strategy in this case, a more detailed analysis might allow for looking at broader trends that exist across U.S. institutions in the technology transfer area, which could then be utilized to guide future reforms.

In terms of the actual reforms, I would also be curious to know more about the types of policies which the authors propose to be implemented in U.S. universities in order to improve the effectiveness of the commercialization pipeline. As part of this question, I recognize the importance mentioning the grassroots student groups who have launched successful programs at MIT and Harvard. Perhaps it would be helpful to suggest how such universities can best share their strategy for success in technology transfer with other universities in the U.S., which could potentially enable the creation of a shared collection of online resources, and/or a network of professionals interested in helping universities improve in this area who would meet regularly to discuss this issue.

Given that there are multiple programs described from these two institutions in Table 2, it may be helpful to discuss more about the specifics of these programs, in terms of the similarities and differences (perhaps in a supplementary figure) in order to point out particular elements that would be useful for other universities to utilize and incorporate into a more universal program that could be adopted widely across the U.S. to improve the technology transfer field. Alternatively, are there differences between these programs that could be pointed out in order to determine their suitability for being adopted by other universities as multiple independent programs?

I also wonder if there are other additional types of actions that could be undertaken by graduate students and postdocs who might have an interest in this area, or by university administrators who could assist in potentially reducing some of these disparities described here. One area that might be worth exploring in this regard is implementing technology transfer programs in all U.S. universities (perhaps these could be designed by a local technology transfer office) geared towards training both researchers and administrators in this area, in order to facilitate the commercialization of the findings produced by researchers in U.S. institutions.

I believe the idea of a third party organization, which is mentioned, could work very well. However, there could be limitations in terms of what universities (or researchers) are willing to share with an outside group, depending on how the agreement terms are formulated. Therefore, it may be useful to also consider other individuals within universities who could be involved in implementing reforms in the technology transfer process from the inside. I also wonder how to ensure that the third party organization will associate with the institutions that are most in need of this partnership, instead of linking to the already high performing institutions, the latter which could in fact ensure their own success. To this end, these agreements would need to be carefully crafted with the interest of both parties in mind.

At the same time, multiple parties should contribute to the implementation of such reforms, including those outside the university. Perhaps it would be useful to add into the current Figure 1 some suggestions at every step along the way, indicating where improvements could be made and by whom (i.e. which stakeholder). This broader analysis might also help with further discussions of the types of changes which should be implemented in universities in order to ensure that all U.S. institutions have the same resources

and as the currently high-performing institutions in technology transfer, therefore aiming to reduce some of these disparities and create a more uniform way to analyze this problem across the board. In addition, as multiple groups are likely to be involved in this process, perhaps it would also be useful to expand this discussion by adding in more details on the efforts already undertaken successfully by each of these groups which could help reduced these disparities. For example this could include direct links to such efforts, with clear instructions and concrete actions which other universities could easily adopt and implement locally.

Specific suggestions:

Overall, the motivation for each part of the study and the actual results are very clearly explained and are insightful. I was particularly intrigued by the findings in Table 1, and the idea that performance of institutions in such these areas can be measured. This is encouraging to see in thinking more broadly about the variables that can be quantified from U.S. institutions. One suggestion is that it may be useful, after each result, or perhaps in the discussion section, to include further explanations of how each of these findings contributes to our broader understanding of academic technology transfer. This might also lead to an obvious list of particular weakness in the commercialization pipeline that must be corrected across institutions.

Getting more into the data itself, with respect to Table 1, I understand the top performing institutions are ranked as high performing in all of the presented categories. However, there are a few institutions in the list that I would consider high performing academically, but which have been indicated to perform in a variable manner in some of the particular areas examined within the commercialization pipeline. I am wondering whether more discussion on these “gray area” institutions could be added, and how their particular performance in a given area translates to benefiting the economy itself. For example, does high performance in certain variables lead to benefits in the local economy, and is high performance in all of these areas required to benefit the U.S. economy as a whole, or is this a more mixed population?

Along these lines, in thinking about the differences in available resources for technology transfer between U.S. universities, I also wonder if it is possible to perform a type of analysis that would take into account this difference and measure performance in a way that is dependent upon the available resources. For example, a particular university may be considered successful overall for their local area given the limited amount of resources, depending on how this analysis is performed and analyzed. However, if we only consider universities that are high performing overall in all categories, it may be more difficult to analyze the individual success of those institutions who only show high performance in certain areas.

Overall, I believe Table 2 contains a wealth of information that could be further analyzed and discussed to draw additional conclusions from the data presented here, including potentially by examining the entire dataset of 300 institutions, which could add another layer of depth and complexity to these findings.

The authors also mentioned that this analysis was performed using 2010-2014 AUTM Licensing Survey data. As this survey data is being reported every year, it may be interesting in the future to expand beyond this timeframe and examine more general trends in the factors presented here over a longer period of time (for example in the last 10 years). That kind of analysis could also provide more insights into the landscape of technology transfer as a whole, and highlight particular recommendations that could be made (or maybe have already been made) to improve the technology transfer system in the U.S. Finally, if available, obtaining and sharing the raw data from the Survey itself may also allow other individuals to perform their own additional analyses of interest.

Is the work clearly and accurately presented and does it cite the current literature?

Yes

Is the study design appropriate and is the work technically sound?

Yes

Are sufficient details of methods and analysis provided to allow replication by others?

Yes

If applicable, is the statistical analysis and its interpretation appropriate?

Yes

Are all the source data underlying the results available to ensure full reproducibility?

Yes

Are the conclusions drawn adequately supported by the results?

Yes

Competing Interests: No competing interests were disclosed.

I have read this submission. I believe that I have an appropriate level of expertise to confirm that it is of an acceptable scientific standard.

Referee Report 19 March 2018

doi:[10.5256/f1000research.15458.r32039](https://doi.org/10.5256/f1000research.15458.r32039)



Karen I. Deak

The IDEA Center, University of Notre Dame, Notre Dame, IN, USA

Other than the Materials and Methods dealing with the underpinnings of the statistical methodology, which are well beyond me (my failure, not the authors'), the article is clear and easy to understand. It provides an interesting look at the University technology transfer process, and is probably of interest to a wide variety of readers -- both those in the technology transfer field and those who care what federal research dollars can provide.

The authors demonstrate that some universities are better than others at commercializing the research that their professors undertake. They provide an examination of whether universities can be good at all steps of the process; or whether it is more effective for a TTO to specialize at one, more high-leverage, point in the process.

The article is clear, and the conclusions and findings are well-reported. The authors additionally provide a few suggestions of how to improve or make more efficient the technology transfer process. These takeaways will be of interest to readers in TTOs, especially.

It would be interesting for a follow-on article for the authors to examine tech transfer efforts at different "types" of universities -- with or without a medical school; or at places where there have been blockbuster licensing deals vs. not. But clearly those ideas are well outside the scope of this article.

Is the work clearly and accurately presented and does it cite the current literature?

Yes

Is the study design appropriate and is the work technically sound?

Yes

Are sufficient details of methods and analysis provided to allow replication by others?

Yes

If applicable, is the statistical analysis and its interpretation appropriate?

I cannot comment. A qualified statistician is required.

Are all the source data underlying the results available to ensure full reproducibility?

Yes

Are the conclusions drawn adequately supported by the results?

Yes

Competing Interests: No competing interests were disclosed.

I have read this submission. I believe that I have an appropriate level of expertise to confirm that it is of an acceptable scientific standard.

The benefits of publishing with F1000Research:

- Your article is published within days, with no editorial bias
- You can publish traditional articles, null/negative results, case reports, data notes and more
- The peer review process is transparent and collaborative
- Your article is indexed in PubMed after passing peer review
- Dedicated customer support at every stage

For pre-submission enquiries, contact research@f1000.com

F1000Research