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Value Added of Universities: Evidence From Georgia

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Abstract

In this paper we propose a quantitative measure of the value added of Georgian universities and their programs, and create a ranking based on that measure. We find that there is a large variation in the rankings of individual programs within universities. At the university level, a comparison with the only available comprehensive ranking reveals that a value added-based ranking can produce very different results from existing popular rankings.

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1. Introduction

For developing countries like Georgia, modernizing and improving the education system is of vital importance for long-term development. Fortunately, there are a number of university choices available to students when applying for undergraduate or graduate studies in Georgia. However, it is not always easy to find an objective measure that can be used to make thoughtful decisions about the relative performance of universities. Moreover, education policy makers in Georgia do not have universal, fair and objective sources upon which they can base decisions on how to distribute funding among public universities. Popular global university rankings often focus only on the top universities and completely neglect most of the universities in poor developing countries that never make it to the top. For example, only one Georgian university is listed in the prestigious Times Higher Education World University Rankings, Tbilisi State University, and it is ranked “1001+”. Needless to say, Georgian universities are not represented in those reputable rankings that cover the top 100 or top 500 universities.

It has long been recognized that many ranking measures suffer from basic flaws. First, the measure of university performance should be implementable and fair for all universities and should not put some institutions in an advantageous position based on factors such as their size alone. Second, the measure of the performance of universities should be focused on *their value added* and should not simply compare student composition. For instance, ranking universities simply based on a comparison of the average entry scores of freshmen students would provide greater indication of how popular the universities are among the best freshmen, rather than of the extent to which those universities *actually improved* their students’ knowledge or skills.

In this paper, we propose a quantitative measure of the value added of Georgian universities and create a ranking based on that measure. By comparing the entry exam scores at the MA level to similar entry scores at the BA level, we derive a value added measure of a particular university and its programs. We then compare our ranking to the only other comprehensive ranking available for Georgian universities: Webometrics (<http://www.webometrics.info/en>). The Webometrics Ranking is produced by the Cybermetrics Lab, a unit of the Spanish National Research Council (CSIC). It is a ranking system based on university web presence, visibility and web access. This ranking system measures how strongly a university is present in the web by its own web domain, sub-pages, rich files, scholarly articles etc. Note that the Webometrics ranks universities and not their programs, so the comparison of the two rankings is possible at the university level only.

We find that the Webometrics ranking favors larger universities, which are not placed at the top of our ranking. Second, we find that universities from outside the capital, Tbilisi, are ranked much higher in our ranking than in the Webometrics ranking. Finally, using a similar methodology we create a ranking of Georgian university undergraduate programs, which to the best of our knowledge is something that has not been done in the academic literature before. We find that there is a large variation in the value added of different programs within universities. This implies that the decisions of university applicants and policy makers should be based on the ranking at the university *program* level and not at the university level. This large variation in the level of value added across different programs provides further motivation to have this detailed ranking be made available.

Simply comparing the MA entry scores of students from different universities or programs in order to rank BA programs potentially suffers from the selection problem. It could be the case that “high-potential” students, those who would improve their skills in all subjects, are characterized by high BA entry scores, or by other observable characteristics such as being a graduate of a better high school. Their improved skills would then mistakenly be attributed to those universities’ BA programs, even though they would have seen a similarly large increase in scores at other universities as well. To combat this problem, we propose a measure of value added that controls for all available observable student characteristics. In this approach, scores in the MA-level entry exam are explained by observable variables including scores in the BA-level entry exam, gender, high school size, location fixed effects, birth year fixed effects, BA admission year fixed effects as well as the duration between BA and MA admissions. The remaining variation in MA entry exam scores across universities (or programs) left unexplained by the other factors is then attributed to the quality of the corresponding universities, i.e. the value added.

The use of value added approaches in the context of evaluating school teachers’ performance is popular in the academic literature. The pioneering work of Hanushek (1971) inspired many other researchers, more recently Rockoff (2004), Rivkin, Hanushek and Kain (2005), Aaronson, Barrow, and Sander (2007), and Chetty, Friedman and Rockof (2014). Although we use a similar approach in the broadest sense, our work differs from these papers as we focus on a developing country and evaluate the value added of universities and their programs. Moreover, while many other studies use experimental data, we use the data from the Georgian universal standardized tests for both undergraduate-level and master’s-level entry exams.

2. Data

Our dataset comes from the Georgian National Assessment and Examination Center (NAEC) and covers the years 2011-2017. It includes information about students’ BA and MA general aptitude entry test results (“entry exam scores”), the years the exams were taken, students’ birth years, BA and MA admission and graduation years, high school location, the gender of the student as well as the size of the high school that the student graduated from.

One of the advantages of our dataset is that it covers the whole population of students taking the unified exams during 2011-2017. Moreover, since the exams are unified – and everyone takes the same exams – the comparison of scores across individuals controlling for student characteristics gives us a sensible measure of the value added. We also compare the general aptitude test results for the BA and MA entry exams, which are similar in nature, and this comparison thus measures the contribution of university undergraduate programs to the development of a student’s general aptitude.

Table 1 summarizes the admissions, total number of universities and corresponding BA programs in Georgia during 2011-2017. As is clear from the table, the number of admissions, total number of universities and number of programs were all increasing over time, but the latter experienced the largest increase. This means that over time relatively more, smaller programs have been introduced. It is thus becoming increasingly important for decision makers to have a measure of university rankings at the program level.

Table 1: University Admissions

| Year | 2011 | 2012 | 2013 | 2014 | 2015 | 2016 | 2017 |
|------------------------|--------|--------|--------|--------|--------|--------|--------|
| Admissions | 23,639 | 26,050 | 28,861 | 26,456 | 28,061 | 27,785 | 28,176 |
| Number of Universities | 58 | 56 | 57 | 60 | 63 | 64 | 65 |
| Number of Programs | 656 | 802 | 1,153 | 1,249 | 1,559 | 1,639 | 1,785 |

In Table 2 we provide the summary statistics and the relationship between the BA score and MA scores for the applicants we have both scores available. The correlation between the two is about 0.7.

Table 2. Summary Statistics of BA and MA Scores

| Variable | Number of Observations | Mean | Standard Deviation | Min | Max |
|-------------------------------------|------------------------|--------|--------------------|------|------|
| General Aptitude Test Result for MA | 11,396 | 7.9 | 3.9 | 0 | 17 |
| General Aptitude Test Result for BA | 11,396 | 1606.5 | 141.1 | 1305 | 1960 |

3. Empirical Specification

In order to control for the observed characteristics of students, we run the following regression:

$$MA_{ij} = \beta_0 + \beta_1 BA_{ik} + \beta_2 Gender_i + \beta_3 S_i + Loc_j + Dur_i + b_y + Ad_y + \varepsilon_{ij} \quad (1)$$

where MA_{ij} denotes the MA entry exam score of student i at university j , BA_{ik} denotes i 's BA entry exam score at university k , $Gender_i$ denotes i 's gender, Dur_i denotes the duration between student i 's BA and MA admissions, b_y , Ad_y and Loc_j denote the Birth Year Fixed Effect, BA Admissions Fixed Effect and Location Fixed Effects, respectively.

The estimated MA entry exam scores, \widehat{MA}_{ij} are then compared to the actual score values MA_{ij} and the differences, i.e. residuals, are taken for each student-university pair. Following the literature, to obtain the measure of university ranking, the average across its students is taken (Jakubowski, 2008; OECD, 2013):

$$VA_j = ave(MA_{ij} - \widehat{MA}_{ij}) = \frac{1}{N_j} \sum_{i=1}^{N_j} \varepsilon_{ij} \quad (2)$$

where N_j is the number of students in university j .

4. Empirical Results

Table 3 shows the regression results where the parameters are estimated using the whole sample of data covering the 2011-2017 entry exam scores at Georgian universities. As can be seen from the table, MA entry exam scores are positively correlated with BA entry exam scores. Moreover, the size of the high school the student graduated from is also positively correlated with their MA entry exam score. However, gender seems to have no significant effect on the exam scores.

Table 5 presents the rank of all Georgian universities with corresponding data: the number of student admissions and number of programs in 2017; the ranking of the best and worst programs of the university, and corresponding student admissions in 2013; and the corresponding Webometrics ranking.

Table 3: Regression Results

| Covariates | Dependent Variable: General Aptitude Test Result for MA Admissions |
|---------------------------------------|--------------------------------------------------------------------|
| BA General Aptitude Test Result | 0.0183*** (0.0002) |
| Gender | 0.056 (0.061) |
| School Size Graduated From | 0.008*** (0.003) |
| Birth Year Fixed Effect | Yes |
| BA Admissions Fixed Effect | Yes |
| Location Fixed Effect | Yes |
| Duration between BA and MA admissions | Yes |
| Observations | 11,396 |
| Adjusted R-squared | 0.461 |

Standard errors in parentheses: *** $p < 0.01$, ** $p < 0.05$, * $p < 0.1$

Dramatic differences can be seen in the two rankings by comparing the first and last columns on Table 5. A couple of observations can be made. First, relative to our value added-based ranking, the Webometrics ranking favors larger universities and those located in the capital. For example, the top three universities in the Webometrics ranking are Tbilisi State University, Ilia State University, and Tbilisi State Medical University with 2017 admissions of 3,409, 3,662 and 1,009 students respectively. All three are based in Tbilisi. However, in our ranking they take 11th, 31st, and 19th places, correspondingly. In contrast, smaller universities (some of which are outside Tbilisi) fare much better in our ranking. For example, the Georgian National Institute Rvali, Zugdidi Teaching University and Sul Khan-Saba Orbeliani Humanities University are in 1st, 3rd and 6th places in our value added-based ranking, while in the Webometrics ranking these institutions do not even make the top 20. As can be seen from Table 3, admissions in 2017 to these universities were much smaller than the admissions to the three universities as ranked top by Webometrics, which tends to favor larger universities. In summary, Table 3 implies that there is almost no relationship between the ranking of Webometrics and our value added-based ranking.

Second, there is a large variation in the rankings of individual programs within universities. Take Gori University, for example. It is placed 16th overall in our ranking, with 227 students admitted

in 2017. However, it has 53 different programs and one of them – *Informatics, Mathematics & Biology* – is the best program (with 13 students admitted in 2013) out of all the different BA programs available in Georgian universities. On the other hand, Gori University’s worst program (which admitted 36 students in 2013) is ranked one of the lowest of all Georgian university programs. Interestingly, the other programs which make the list include, for example, increasingly attractive programs in Georgia such as *Tourism, Business Administration, Management, and Agricultural Studies*.

Table 4: Program Ranking

| Program Rank | University | Category |
|--------------|---------------------------------------------------------------|------------------------------------|
| 1 | Gori University | Informatics, Mathematics & Biology |
| 2 | International Black Sea University | Tourism |
| 3 | Euroregional University | Psychology |
| 4 | David Guramishvili International Teaching University "iberia" | Business Administration |
| 5 | Shota Rustaveli Theater and Film University | Management |
| 6 | Ilia Chavchavadze Georgian National University | Journalism |
| 7 | Akaki Tsereteli State University Kutaisi | Design |
| 8 | Caucasus University | Cinematography |
| 9 | Georgian Technical University | Geodesy |
| 10 | University of Georgia | Health Administration |
| 11 | Telavi Iakob Gogebashvili State University | Agricultural Studies |
| 12 | David Aghmashenebeli University of Georgia | Health Care |
| 13 | Ilia State University | Sports |
| 14 | Grigol Robakidze University Tbilisi | Journalism |
| 15 | Akaki Tsereteli State University Kutaisi | Agricultural Studies |
| 16 | Zugdidi Teaching University (Zugdidi) | Business Administration |
| 17 | Georgian National Institute Rvali | Accounting |
| 18 | Ilia Chavchavadze Georgian National University | Journalism |
| 19 | Ilia State University | Law |
| 20 | Caucasus University | Humanitarian Studies |

5. Conclusion

In this paper we examined entry exam scores at Georgian universities. Controlling for students’ characteristics, we constructed a measure of value added of the universities and their programs.

We find that a value added-based ranking is very different from the only other ranking available for Georgian universities. We also find that the programs within universities are ranked very differently. Perhaps students should take this into account when making decisions about their education.

The recent papers on the related topics demonstrate the significance of the research on education attainments in the developing countries. For example, using the recent Skills Towards Employability and Productivity (STEP) surveys of urban labor force participants Shafiq et al. (2018) examine individuals' educational attainment, labor market participation, and earnings. Using logistic regressions, they find that individuals from disadvantaged origins are less likely to obtain a higher education degree. For future work, it is interesting to explore which universities or programs are better at improving the social mobility of students who do get the chance to obtain a higher education degree.

6. References

1. Aaronson, Daniel, Lisa Barrow, and William Sander (2007). "Teachers and student achievement in the Chicago public high schools." *Journal of Labor Economics* 25, no. 1: 95-135.
2. Chetty, Raj, John N. Friedman, and Jonah E. Rockoff (2014). "Measuring the impacts of teachers II: Teacher value-added and student outcomes in adulthood." *American Economic Review* 104, no. 9: 2633-79.
3. Hanushek, Eric (1971). "Teacher characteristics and gains in student achievement: Estimation using micro data." *The American Economic Review* 61, no. 2: 280-288.
4. Jakubowski, Maciej (2008). Implementing value-added models of school assessment. *EUI Working Papers RSCAS 2008/06, European University Institute*.
5. Kim, HoonHo, and Diane Lalancette (2013). "Literature review on the value-added measurement in higher education." *Paris, France: OECD*.
6. Rivkin, Steven G., Eric A. Hanushek, and John F. Kain (2005). "Teachers, schools, and academic achievement." *Econometrica* 73, no. 2: 417-458.
7. Rockoff, Jonah E (2004). "The impact of individual teachers on student achievement: Evidence from panel data." *American Economic Review* 94, no. 2: 247-252.
8. Shafiq, M. Najeeb, Robert Toutkoushian and Alexandria Valerio (2019). Who benefits from higher education in low- and middle-income countries? *Journal of Development Studies* 55(11), pp. 2403-2423.

7. The Main Table
Table 5: University and Program Ranking

| Rank | University Name | University Admissions in 2017 | Number of Programs in 2017 | Rank of the Top Program | Admissions at the Top Program in 2013 | Rank of the Worst Program | Admissions at the Worst Program in 2013 | Webometrics Ranking | Point | CI |
|------|---------------------------------------------------------------|-------------------------------|----------------------------|-------------------------|---------------------------------------|---------------------------|-----------------------------------------|---------------------|--------|------------------|
| 1 | Georgian National Institute Rvalli | 14 | 7 | 18 | 30 | 775 | 63 | 41 | 1.650 | (1.007, 2.293) |
| 2 | Tbilisi University | 27 | 13 | 112 | 14 | 120 | 89 | 50 | 1.306 | (0.717, 1.896) |
| 3 | Zugdidi Teaching University (Zugdidi) | 36* | 16* | 16 | 6 | 751 | 11 | 28 | 1.091 | (0.642, 1.539) |
| 4 | University Interpharm+ | 48* | 2* | 146 | 47 | 146 | 47 | 56 | 1.055 | (-0.296, 2.405) |
| 5 | V. Saradjishvili Tbilisi State Conservatoire | 77 | 19 | 155 | 3 | 155 | 3 | 17 | 1.014 | (-1.318, 3.346) |
| 6 | Sul Khan-Saba Orbeliani Humanities University | 186 | 23 | 86 | 57 | 241 | 91 | 33 | 0.911 | (0.437, 1.386) |
| 7 | Tbilisi Teaching University "gorgasali" | 27 | 11 | 25 | 16 | 697 | 47 | N/A | 0.851 | (0.563, 1.139) |
| 8 | High School Georgia | 7 | 3 | 124 | 15 | 349 | 69 | N/A | 0.704 | (0.283, 1.125) |
| 9 | Akhalqalaqi High School-College (Akhalqalaqi) | 16 | 3 | 157 | 16 | 352 | 3 | N/A | 0.695 | (-0.045, 1.434) |
| 10 | Free University of Tbilisi | 528 | 23 | 119 | 577 | 535 | 100 | 9 | 0.665 | (0.633, 0.698) |
| 11 | Ivane Javakishvili Tbilisi State University | 3409 | 110 | 44 | 55 | 685 | 47 | 1 | 0.292 | (0.291, 0.294) |
| 12 | Sukhishvili Teaching University (Gori) | 35 | 16 | 78 | 3 | 784 | 3 | N/A | 0.290 | (0.21, 0.37) |
| 13 | Saint Tbel Abuseridze University (Khulo) | 37 | 11 | 107 | 94 | 687 | 3 | N/A | 0.153 | (0.128, 0.178) |
| 14 | Caucasus International University | 483 | 27 | 58 | 163 | 729 | 75 | 19 | 0.135 | (0.107, 0.164) |
| 15 | David Guramishvili International Teaching University "iberia" | 271* | 11* | 4 | 12 | 795 | 18 | N/A | 0.124 | (-0.009, 0.258) |
| 16 | Gori University | 227 | 53 | 1 | 13 | 797 | 36 | 25 | 0.106 | (0.063, 0.148) |
| 17 | Tbilisi David Aghmashenebeli University | 68 | 8 | 24 | 48 | 782 | 12 | 47 | 0.083 | (-0.008, 0.174) |
| 18 | Georgia State Agriculture University | 420 | 19 | 35 | 25 | 718 | 20 | 7 | 0.063 | (0.029, 0.097) |
| 19 | Tbilisi State Medical University | 1009 | 31 | 47 | 68 | 783 | 22 | 3 | 0.051 | (0.032, 0.069) |
| 20 | University of Georgia | 1108 | 86 | 10 | 36 | 762 | 9 | 6 | 0.028 | (0.016, 0.04) |
| 21 | Sokhumi State University | 643 | 38 | 67 | 98 | 660 | 30 | 16 | 0.013 | (-0.009, 0.036) |
| 22 | Batumi Shota Rustaveli State University | 1241 | 82 | 28 | 4 | 802 | 3 | 5 | -0.006 | (-0.014, 0.002) |
| 23 | David Aghmashenebeli University of Georgia | 271 | 27 | 12 | 337 | 757 | 9 | 37 | -0.024 | (-0.059, 0.011) |
| 24 | Akaki Tsereteli State University Kutaisi | 1611 | 91 | 7 | 2 | 778 | 20 | 11 | -0.070 | (-0.077, -0.062) |
| 25 | Guram Tavartkiladze Teaching University | 99 | 16 | 173 | 112 | 727 | 76 | 35 | -0.076 | (-0.65, 0.497) |
| 26 | Caucasus Academic Center (CAC) | 79* | 7* | 39 | 17 | 789 | 4 | N/A | -0.081 | (-0.461, 0.298) |
| 27 | Caucasus University | 961 | 89 | 8 | 60 | 759 | 11 | 8 | -0.117 | (-0.146, -0.089) |
| 28 | International Black Sea University | 531 | 76 | 2 | 18 | 764 | 44 | 4 | -0.116 | (-0.157, -0.09) |
| 29 | American University for Humanities | 21* | 2* | 323 | 7 | 459 | 79 | 34 | -0.123 | (-0.542, 0.292) |
| 30 | Telavi Iakob Gogebashvili State University | 270 | 41 | 11 | 7 | 801 | 4 | 20 | -0.171 | (-0.226, -0.115) |
| 31 | Ilia State University | 3662 | 93 | 13 | 1 | 796 | 14 | 2 | -0.183 | (-0.184, -0.177) |
| 32 | Shota Rustaveli Theater and Film University | 132 | 11 | 5 | 22 | 739 | 25 | 23 | -0.196 | (-0.281, -0.114) |
| 33 | University of Tsodna | 140* | 5* | 445 | 83 | 449 | 37 | 43 | -0.208 | (-0.62, 0.191) |
| 34 | University Geomedi | 162 | 13 | 126 | 105 | 786 | 6 | 38 | -0.214 | (-0.36, -0.079) |
| 35 | Georgian Technical University | 4320 | 201 | 9 | 4 | 803 | 10 | 10 | -0.226 | (-0.23, -0.221) |
| 36 | Tbilisi Humanities University | 38 | 9 | 141 | 29 | 624 | 72 | 36 | -0.230 | (-0.4, -0.069) |
| 37 | Batumi Arts Teaching University | 330 | 26 | 42 | 100 | 792 | 7 | 30 | -0.235 | (-0.29, -0.181) |
| 38 | Kutaisi University | 23 | 10 | 118 | 62 | 749 | 17 | 29 | -0.290 | (-0.429, -0.15) |

| | | | | | | | | | | |
|----|--------------------------------------------------------------|------|----|-----|-----|-----|-----|-----|--------|------------------|
| 39 | Grigol Robakidze University Tbilisi | 325 | 42 | 14 | 17 | 785 | 1 | 13 | -0.371 | (-0.412, -0.342) |
| 40 | Georgian American University | 251 | 32 | 49 | 197 | 589 | 212 | 22 | -0.380 | (-0.422, -0.338) |
| 41 | Shota Meskhia Zugdidi State Teaching University | 95 | 22 | 235 | 18 | 798 | 12 | 28 | -0.432 | (-0.483, -0.387) |
| 42 | Georgian University of Saint Andrew | 287 | 34 | 22 | 7 | 790 | 8 | 14 | -0.467 | (-0.746, -0.189) |
| 43 | Zurab Zhvania Georgian Institute of Public Affairs | 192 | 19 | 421 | 185 | 673 | 136 | 12 | -0.486 | (-0.649, -0.326) |
| 44 | Business Academy of Georgia | 575 | 18 | 318 | 585 | 743 | 449 | 48 | -0.503 | (-0.644, -0.338) |
| 45 | Samtskhe Javakheti State University | 305 | 33 | 143 | 4 | 777 | 12 | 46 | -0.508 | (-0.546, -0.465) |
| 46 | Ilia Chavchavadze Georgian National University | 2057 | 58 | 6 | 9 | 678 | 459 | 32 | -0.582 | (-0.596, -0.56) |
| 47 | Rustavi College | 37* | 5* | 452 | 35 | 708 | 2 | 59 | -0.588 | (-1.403, 0.217) |
| 48 | David Aghmashenebeli Defence Academy of Georgia | 71 | 4 | 21 | 11 | 765 | 428 | 44 | -0.601 | (-0.724, -0.483) |
| 49 | Saint Queen Tamar Teaching University | 16 | 9 | 135 | 26 | 735 | 1 | N/A | -0.609 | (-1.224, 0.006) |
| 50 | Tbilisi Open Teaching University | 86 | 21 | 97 | 114 | 787 | 67 | N/A | -0.645 | (-0.713, -0.573) |
| 51 | Euroregional University | 50* | 4* | 3 | 14 | 800 | 29 | 60 | -0.844 | (-1.371, -0.323) |
| 52 | Batumi State Maritime Academy | 44 | 6 | 130 | 8 | 763 | 18 | 15 | -0.854 | (-1.407, -0.3) |
| 53 | Georgian Aviation University | 172 | 30 | 440 | 208 | 793 | 44 | 27 | -0.950 | (-1.085, -0.793) |
| 54 | David Tvildiani Medical University (AIETI Medical School) | 120 | 7 | 635 | 380 | 635 | 380 | 26 | -0.990 | (-4.508, 2.541) |

Notes: *admission and programs information are not available for these universities in 2017, so the data for the most recent years are taken. Some of the universities are not ranked by Webometrics and their rankings are denoted by "N/A". The last two columns indicate the point estimates and 95% confidence intervals of these estimates.