

Investment in High Performance Computing A Predictor of Research Competitiveness in U.S. Academic Institutions

Amy Apon, Ph.D.

Director, Arkansas High Performance Computing Center Professor, CSCE, University of Arkansas

Stan Ahalt, Ph.D.

Director RENCI Professor, Computer Science, UNC-CH Work supported by the NSF through Grant #0946726 University of Arkansas and RENCI/UNC-CH



Collaborators

Amy Apon University of Arkansas Stanley Ahalt RENCI, University of North Carolina Vijay Dantuluri RENCI, University of North Carolina Constantin Gurdgiev IBM

Moez Limayem University of Arkansas Linh Ngo University of Arkansas Michael Stealey RENCI, University of North Carolina

Research Study

- Background and motivation
- Research hypothesis
- Data acquisition
- Analysis and Results
- Discussion

Research and Computing



Computational and Data Driven Science

Cyberinfrastructure Ecosystem Foundation

CyberInfrastructure Ecosystem



Conversation with a Chancellor

 HPC guys, "This is a great investment! We think we can run the HPC center with only \$1M/year in hardware and \$1M/year in staffing."



Chancellor, "Which 20 faculty do you want me to fire?"

HPC: High rePeating Cost

- Computer equipment is usually treated as a capital expense, with costs for substantial clusters in the range of \$1M+
- Warranties on these generally last 3 years, or 5 years at most, after which repairs become prohibitive
- Even without that, the pace of technology advances require refreshing every 3-5 years
- Staffing is a long term repeating cost!



Ranks of Top 500 Computers and Appearances in Succeeding Lists

Some Observations

Tflops versus Core Hours Used

Academic HPC Centers



What is the ROI?

 Can I convince my VPR that the funds invested in HPC add value to the institution and create opportunity?

What if this is not true?

Hypothesis

 Investment in high performance computing, as measured by entries on the Top 500 list, is a predictive factor in the research competitiveness of U.S. academic institutions.

We study Carnegie Foundation institutions with "Very High" and "High" research activity – about 200 institutions

Data Acquisition

Independent variables

- Top 500 List count and rank of entries
 - Mapped from "supercomputer site" to "institution"
 - We note that entries are voluntary the absence of an entry does not mean that an institution does not have HPC

Dependent variables

- NSF and other federal funding summary and award information
- Publication counts
- U.S. News and World Report rankings

Data from the Top 500 List



An historical record without comparison of supercomputers



institutions as they appear cumulatively

no. of academic institutions

----no. of machine entries

About 100 U.S. institutions have appeared on a Top 500 List



- Examples
- Correlation analysis
- Regression analysis

Simple Example of ROI

Evidence based on 2006 NSF funding

With HPC

Without HPC



Longer Example of ROI

More evidence, 1993-2009 NSF funding



Correlation Analysis

	Counts	NSF	Pubs	All Fed	DOE	DOD	NIH	USNews
dRankSum	0.8198	0.6545	0.2643	0.2566	0.2339	0.1418	0.1194	-0.243
Counts		0.6746	0.4088	0.3601	0.3486	0.1931	0.2022	-0.339
NSF			0.7123	0.6542	0.5439	0.2685	0.4830	-0.540
Pubs				0.8665	0.4846	0.3960	0.8218	-0.588
All Fed					0.4695	0.6836	0.9149	-0.543
DOE						0.1959	0.3763	-0.384
DOD							0.4691	-0.252
NIH								-0.500

Regression Analysis

- Two Stage Least Squares (2SLS) regression is used to analyze the research-related returns to investment in HPC
- We model two relationships
 - Model 1: NSF Funding as a function of contemporaneous and lagged Appearance (APP) on the Top 500 List Count and Publication Count (PuC), and
 - Model 2: Publication Count (PuC) as a function of contemporaneous and lagged Appearance on the Top 500 List Count (APP) and NSF Funding

Endogeneity

- Funding allows an institution to acquire resources
- Resources are used to perform research, which leads to more funding
- Resources are also cited in the argument for research funding
- NSF funding begats HPC resources which begats NSF funding ...

Regression Analysis

- Original tests revealed significant problems with endogeneity of Publication Counts (PuC) and NSF Funding.
- To correct for this, we deployed a 2SLS estimation method, with number of undergraduate Student Enrollments (SN) acting as an instrumental variable in the first stage regression for PuC (Model 1) and NSF (Model 2).
- In both cases, SN was found to be a suitable instrument for endogenous regressors.

First Result

- <u>A single HPC investment yields</u> statistically significant immediate returns in terms of <u>new NSF funding</u>
- An entry on a list results in an increase of yearly NSF funding of \$2.4M

o Confidence level 95%
 o Confidence interval \$769K-\$4M

Second Result

- <u>A single HPC investment yields</u> statistically significant immediate returns in terms of <u>increased</u> <u>academic publications</u>
- An entry results in an increase in yearly publications of 60

 Confidence level 95%
 Confidence interval 19-100

Third Result

- Analysis on the rank of the system shows that rank has a positive impact to competiveness, but with reduced confidence.
- We have not studied returns to other institutions of investments by resource providers, or returns to overall U.S. competitiveness.

Fourth Result

- HPC investments suffer from fast depreciation over a 2 year horizon
- Consistent investments in HPC, even at modest levels, are strongly correlated to research competitiveness.
- Inconsistent investments have a significantly less positive ROI

Discussion

- More study is needed to precisely determine the rate of depreciation of HPC investments
- The publication counts include all publications, not just those related to HPC
- More study is needed regarding how use of national systems, such as Teragrid, may impact research competitiveness

Data from Teragrid Usage



Questions?