

Predicting Job Placements of Economics Graduate Programs

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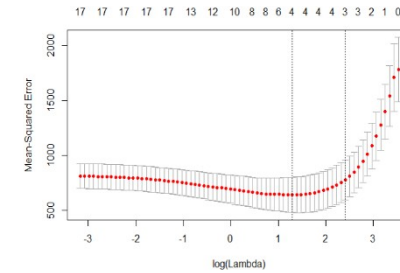
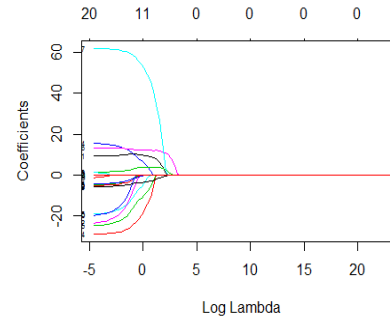
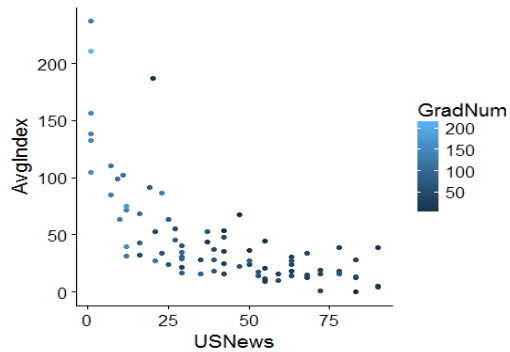
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Outline



Data

- Program Characteristics
- Placement Quality Measure

Model

- Lasso
- Cross Validation

Result

- Implication
- Future Work

Literature on “New Economists”

Paper	Authors	Data	Method	Result (Predictors)
What Does Performance in Graduate School Predict? (2007)	Susan Athey, Lawrence F. Katz, Alan B. Krueger, Steven Levitt, and James Poterba	1029 grad students at Top 5 in the 1990s	Descriptive Statistics, OLS	First-year course grades, Top undergraduate institutions
Forecasting Job Placements of Economics Graduate Students (2000)	Alan Krueger and Stephen Wu	344 students in one department in 1989	Tobit model and OLS	GRE (math and economics), ratings of the admissions committee
15 Years of Research on Graduate Education in Economics (2014)	John Siegfried and Wendy Stock	585 students at 27 programs	Descriptive Statistics, OLS	Office spaces and research assistantships lead to higher completion rate
Predicting Job Placements of Economics Graduate Programs	(In Progress)	104 programs' average placement quality (covering 3915 graduates) from 2010 to 2017	Data Visualization, Regularized Regression, Model Selection	Faculty Productivity

Program Characteristics ("X")



- The National Research Council's "A Data-Based Assessment of Research-Doctorate Programs" in 2010
- Examines the program characteristics of 110 Economics Ph.D. doctoral programs in the U.S
- Private or public universities, location, scholarship amount, size, providing office space or not, diversity, the number of publications per faculty member, funding, citations per publication, number of grants, diversity, etc.
- We use 18 program-value exogenous variables that are related to students' placement in the following analysis.

[illegible]

- Compiled the initial job placement of these 110 Economics doctoral programs from 2010 to 2017.
- No data on Five programs
(Auburn, Northern Illinois, RPI, SUNY Buffalo and Utah State)
- 3915 observations specifying each student's pedigree, graduation year, initial employer and corresponding positions
- Ex:
University of A (2015) → University of B, Assistant Professor

Choose the ranking

Name	Coverage	Methodology	Latest Data
U.S News Economics Ranking	Top 90 U.S Programs	Results of peer assessment surveys	2017
“Ranking of Academic Journals and Institutions in Economics” (Kalaitzidakis et al., 2003)	Top 200 programs	Publications in 30 leading economics journals	2003
RePEc/IDEAS rankings	Top 10% of Institutions (Scores for 378 programs)	Publications of registered authors	2017
“Ranking Economics Departments Worldwide on the Basis of PhD Placement” (Amir and Knauff, 2005)	Top 58 worldwide departments	Sum of the values of its PhD graduates	2006
The Tilburg University Worldwide Economics Schools Research Ranking	About 1000 worldwide programs with at least one publications from 2010 to 2017	Publications in 70 leading economics journals	2017

Matching Placement With Tilburg Ranking

- ❖ The journal counts are treated as the scores to indicate the placement quality.
- ❖ There are 275 departments tied with rank 988 for $s=1$ (one publication)
- ❖ 3003 out of 3915 placements (77%) are successfully assigned scores
- ❖ The remaining 912 entries are assigned 0.
- ❖ Discount rate for different positions (Krueger and Wu (2000)))

Job Positions	Discount Rate	Counts(%)	Comment
Tenure-track assistant professor	1	3231 (83%)	Most ideal academic placement
Postdocs	0.6	520 (13%)	Pre-assistant professors
Lecturer/Visiting/Practice/Clinical assistant professor	0.4	130 (3%)	Non-tenure track, teaching positions

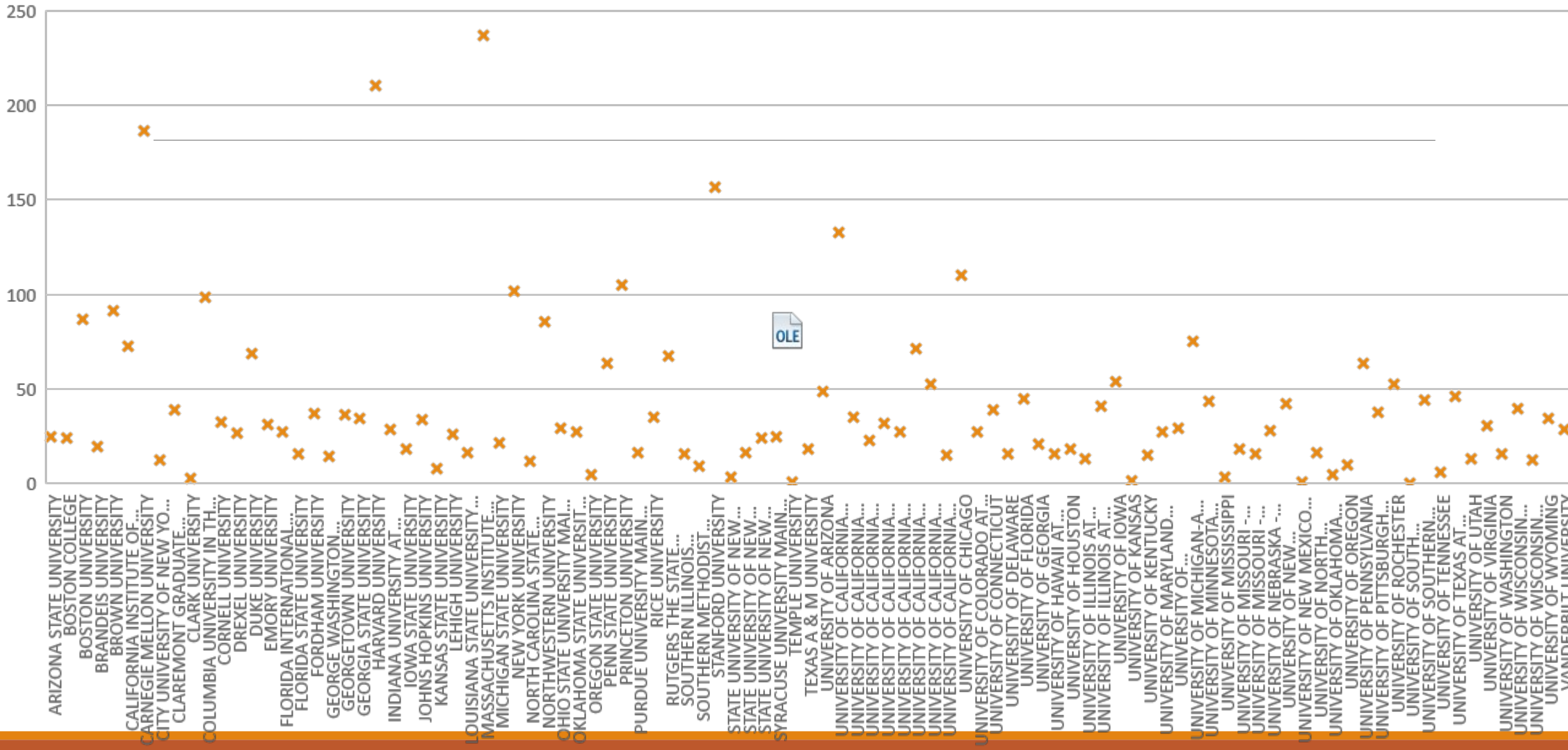
AvgIndex("y")

- AvgIndex = $\frac{\text{sum of all graduates' placement scores}}{\text{number of graduates}}$

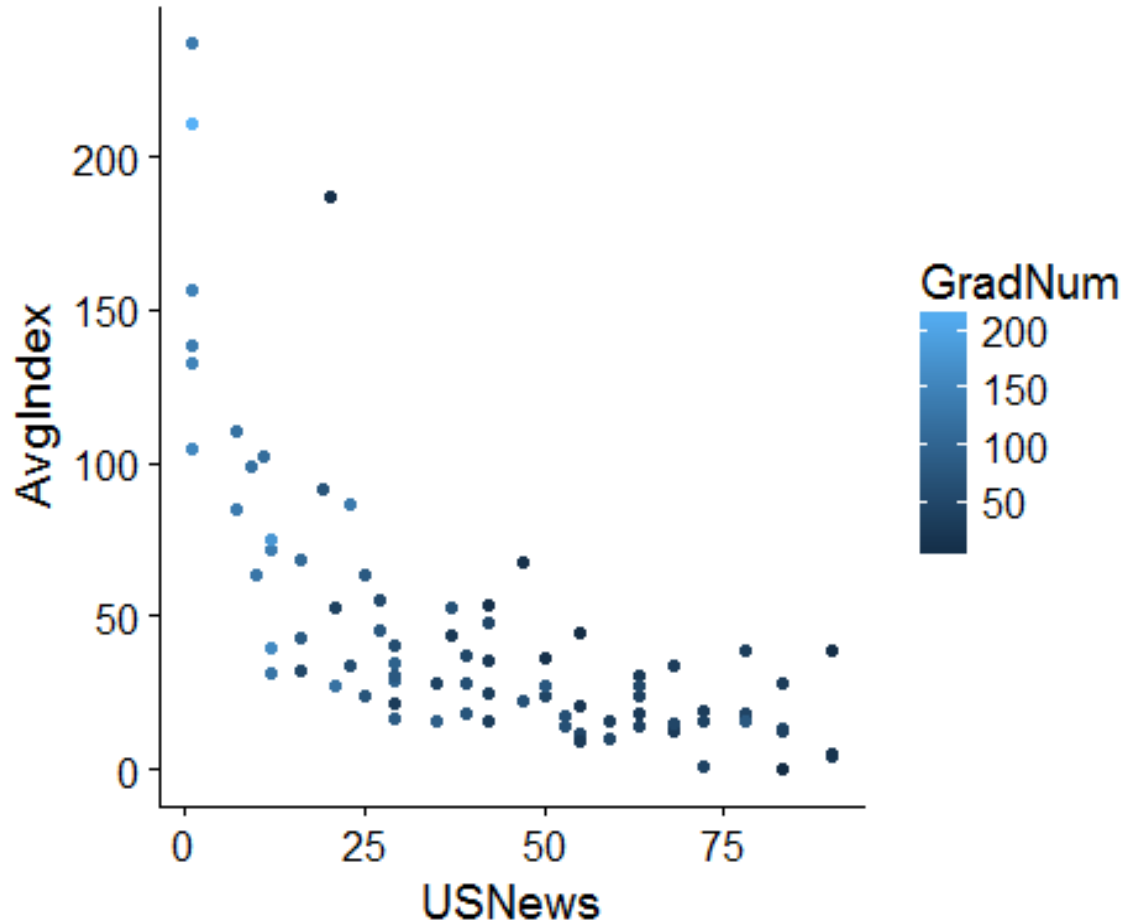
Quantiles	AvgIndex	Example Institutions
0%	0	Gettysburg College
25%	15	Miami University
50%	28	Fordham University
75%	45	University of Georgia
100%	237	Cornell University

Visualization of AvgIndex

Series1 Series2 Series3

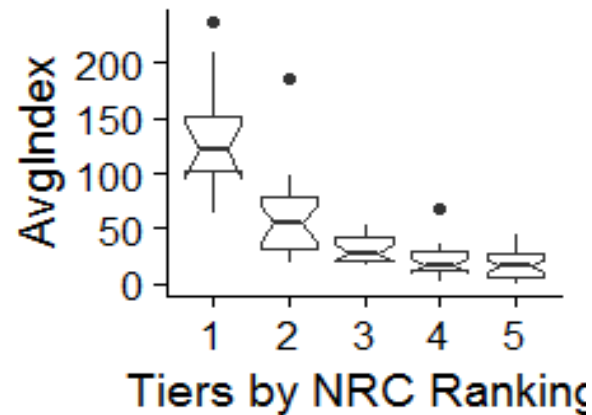
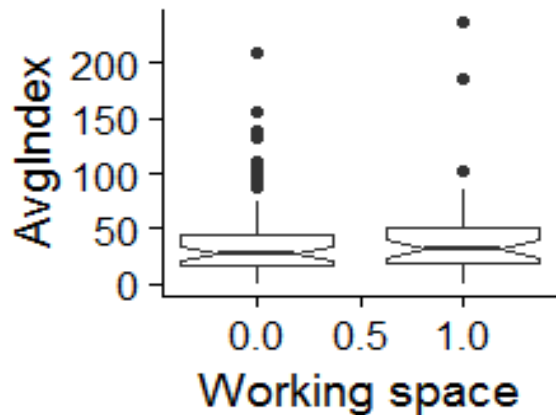
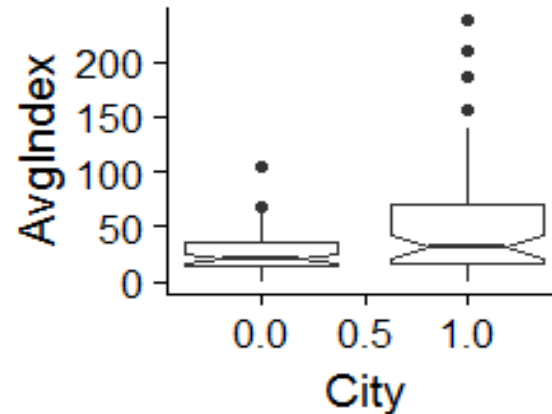
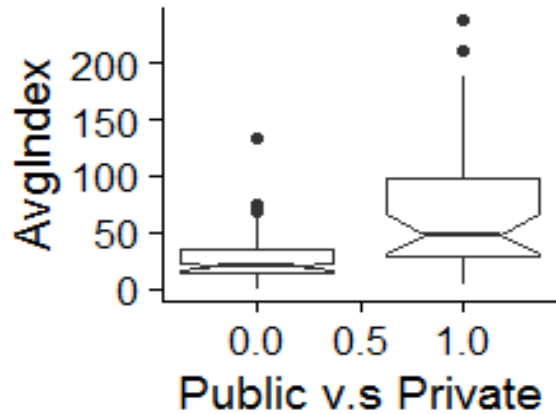


Does US News Ranking predict “y”?



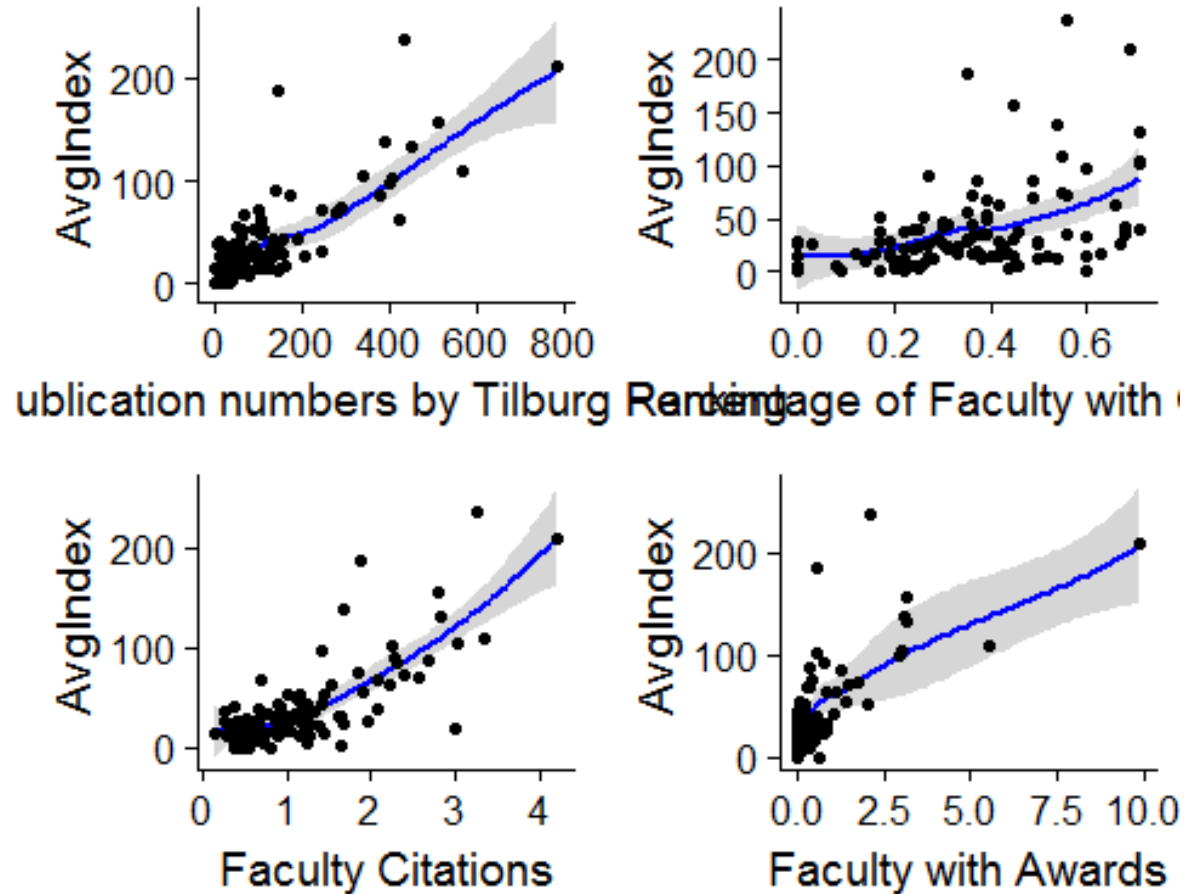
- ❑ Plot the 'AvgIndex' as a function of 'USNews'
- ❑ The size of dot : relative number of graduates
- ❑ The AvgIndex increases when the rank climbs
- ❑ Among Top 25 programs, AvgIndex increases more steeply when the rank goes up.
- ❑ There are no significant AvgIndex differences among the rest of the programs.

Explore “X” (categorical) and “y”



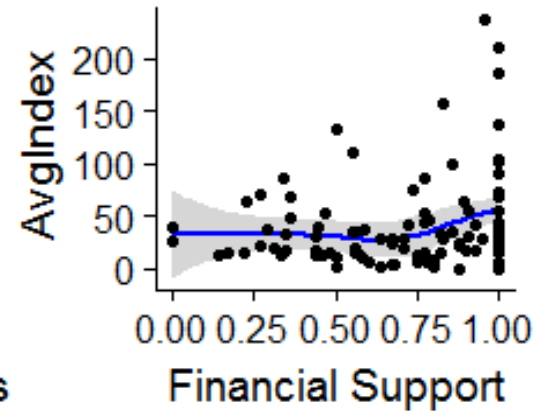
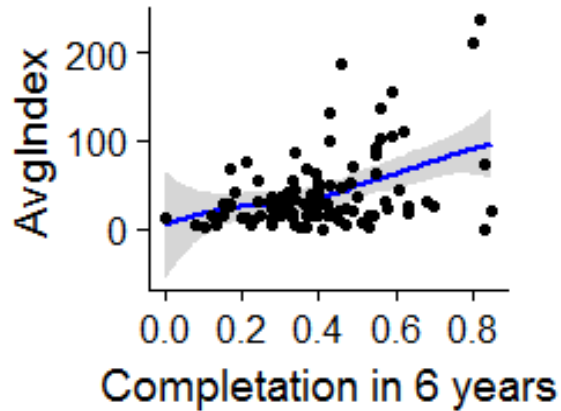
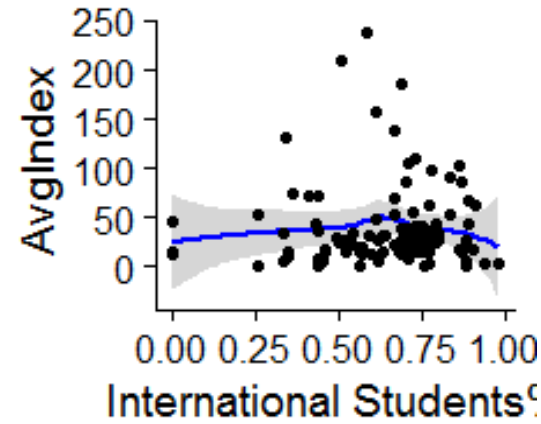
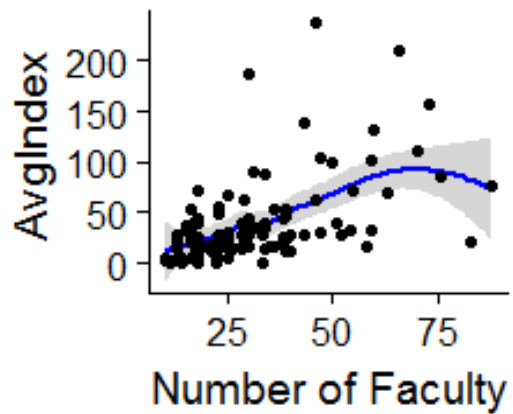
- ❖ Private institutions have higher AvgIndex on average
- ❖ No significant difference caused by campus setting. Best performing departments are in a city with 10k population.
- ❖ Working Space has no effect on AvgIndex
(Stock and Siegfried (2006) found providing working space increased completion rate)
- ❖ The differences among Top 30-104 are small

Explore “X” (continuous) and “y”



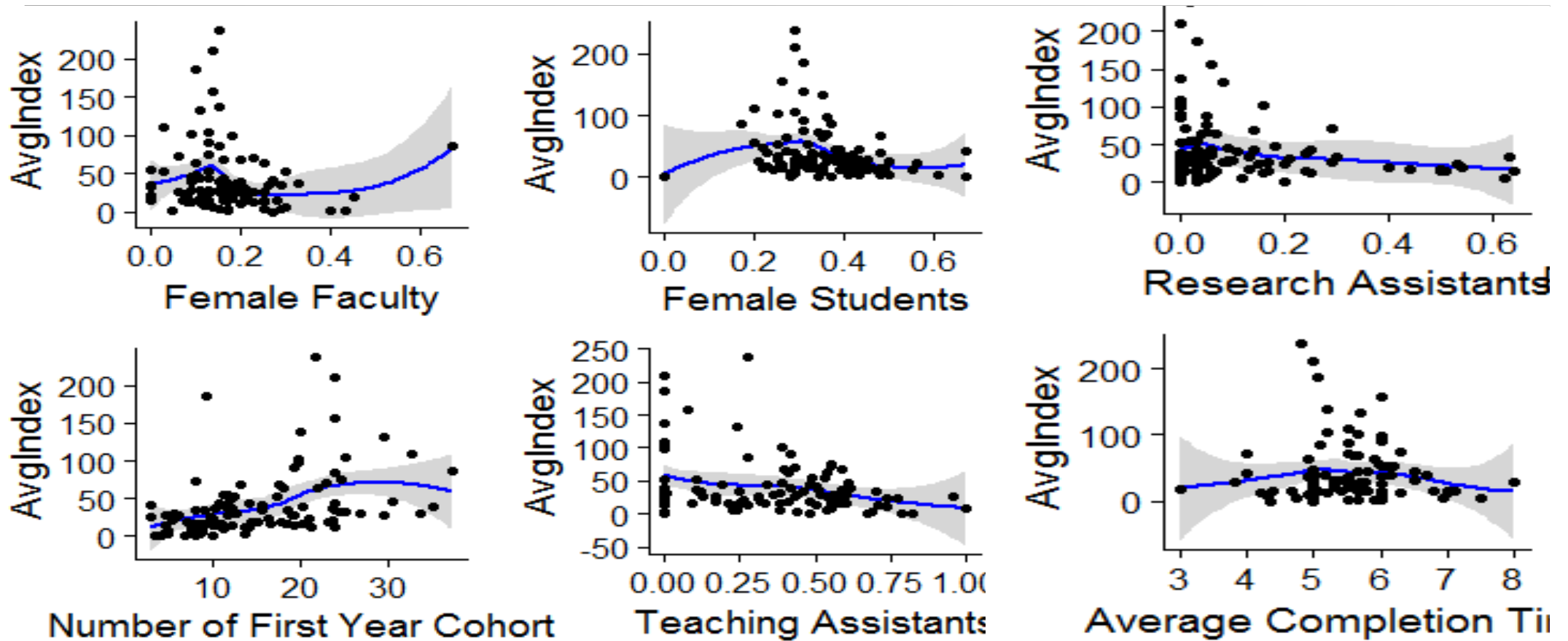
- ❖ Publication numbers (Tilburg), Grants, Citations, Awards are indicators of faculty quality
- ❖ They are highly correlated with AvgIndex

Explore “X” (continuous) and “y”



- ❖ Financial Support has no significant effect on AvgIndex
- ❖ Bolli et al. (2015) found institutional student support caused higher completion rates

Explore “X” (continuous) and “y”



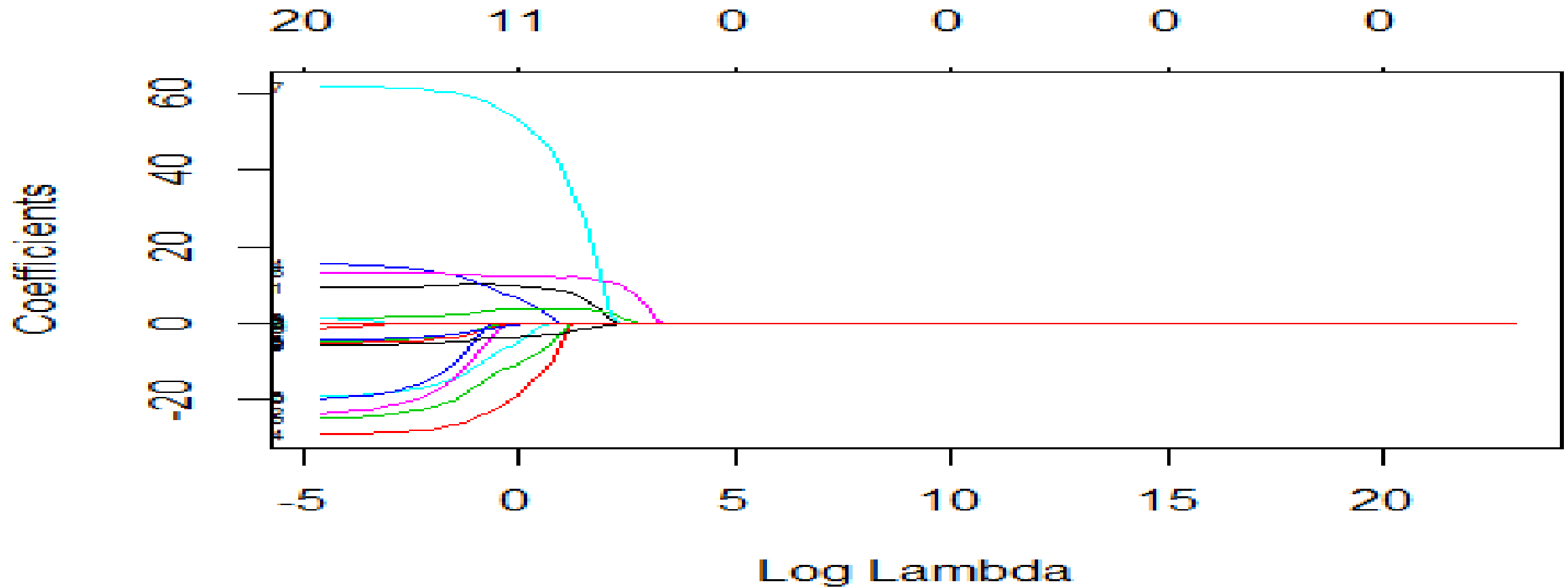
Regularized Regression

- We use the Lasso to fit the prediction model
- Reasons: 1. Large number of predictors (18)
 2. Prefer a model with smaller variance and MSE (OLS gives high variance)
 3. More interpretable model (Ridge uses l-2 in penalty term; keeps all coefficients)
- Coefficients minimize the quantity

$$\sum_{i=1}^n \left(y_i - \beta_0 - \sum_{j=1}^p \beta_j x_{ij} \right)^2 + \lambda \sum_{j=1}^p |\beta_j| = \text{RSS} + \lambda \sum_{j=1}^p |\beta_j|.$$

- In the penalty term, the tuning parameter λ controls the degree of penalty.
- If λ becomes larger, the penalty term will shrink more coefficients to be zero.
- If λ is 0 (unregularized), we get same coefficients from least square fit.

For each value of λ , we get different sets of coefficient estimates



Model Selection

- Cross-validation helps us estimate prediction error and select optimal model

- Step 1: Split the data into 5 folds. Start with initial $\log(\lambda)$



- Step 2: Leave out part 1, fit the model to the rest;

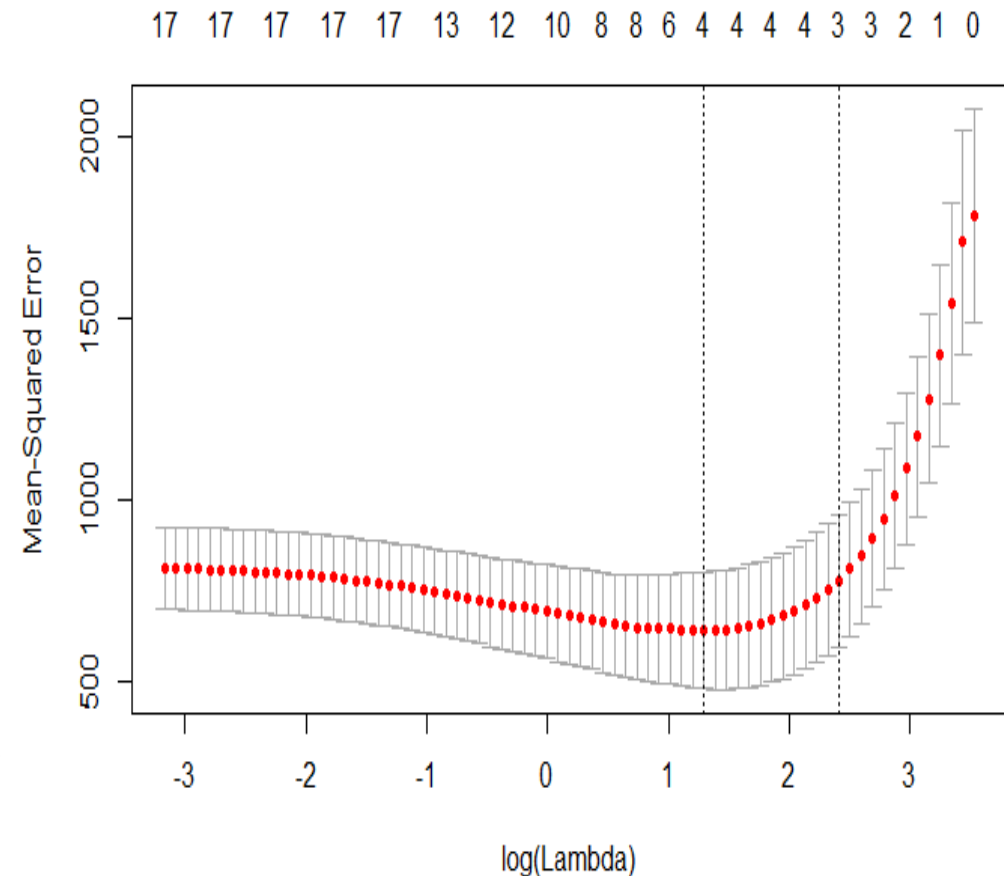
Obtain MSE for the 1st part

- Step 3: Repeat the process for $k=2,3,4,5$

$$CV_{(K)} = \sum_{k=1}^K \frac{n_k}{n} MSE_k$$

where $MSE_k = \sum_{i \in C_k} (y_i - \hat{y}_i)^2 / n_k$, and \hat{y}_i is the fit for observation i , obtained from the data with part k removed.

- Step 4: Iterate for different values of $\log(\lambda)$
- Step 4: Iterate for different values of)



Model with minimum MSE

(Intercept)	5.9870935
Pri	7.2644151
City	.
Space	.
FS	.
International	.
Citations	12.5342821
Faculty	.
Grants	.
Awards	4.3116645
`Completion%`	.
`Female Faculty`	.
`Female Students`	.
`First Year`	.
RA	.
TA	.
Time	.
GradNum	.
Tilburg	0.1196028

3.03 (Four predictors)

Useful for predicting updated AvgIndex on new NRC data

For department chairs and administrators,

hiring productive professors may be the best strategy to boost placement quality

For graduate students (current and potential), potential advisors' research productivities matter a lot if they want to start a promising career in academia.

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