# Predicting Not-for-Profit Executive Compensation

A data science project using classification to model executive salaries as reported on IRS form 990

#### Form 990 Return of Organization Exempt From Income Tax Under section 501(c), 527, or 4947(a)(1) of the Internal Revenue Code (except black lung benefit trust or private foundation) The organization may have to use a copy of this return to satisfy state reporting requirements. A For the 2005 calendar year, or tax year beginning 2005, and ending D Employer ident B Check if applicable: Address change E Telephone nun Name change Initial return City or town, state or country, and ZIP + 4 F Accounting method: Final return Amended return H and I are not applicable to sect Application pending • Section 501(c)(3) organizations and 4947(a)(1) nonexempt charitable trusts must attach a completed Schedule A (Form 990 or 990-FZ). H(b) If "Yes," enter number of aff G Website: ▶ J Organization type (check only one) ► ☐ 501(c) ( ) ◄ (insert no.) ☐ 4947(a)(1) or ☐ 527 H(d) Is this a separate return filed by K Check here ▶ ☐ if the organization's gross receipts are normally not more than \$25,000. The organization covered by a group organization need not file a return with the IRS; but if the organization chooses to file a return, be

### Scope & Interest in This Topic

1.5M not-for-profit organizations in the US advocate for causes that are highly sympathetic. These organizations easily pull at the heartstrings of donors. But how is that money being spent?

My goal with this project was to design a tool that could be used to promote not-for-profit accountability and fair compensation.

<u>Charity Navigator</u> has used a numbers-based rating system to score over 9,000 not-for-profits. A model that can identify outliers could be used to flag specific organizations that are misusing donations.

Additionally, the Board of Directors of a not-for-profit could use this tool to determine a fair salary based on the resources and nuances of its not-for-profit.







### **Data Source**





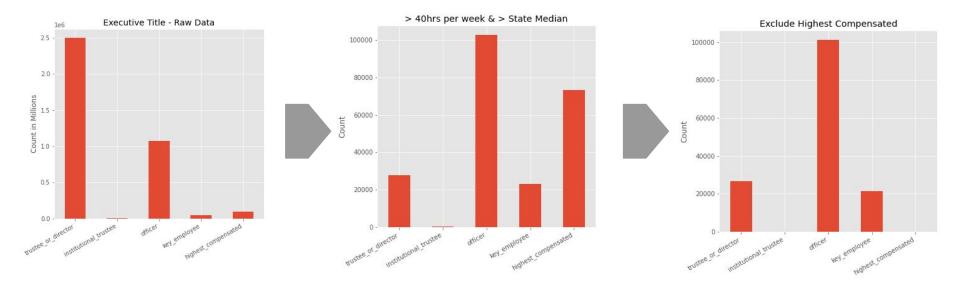
First <u>made available</u> by the IRS in 2016 and hosted by both <u>Amazon Web</u>
<u>Services</u> (xml format) and <u>Open990</u> (csv format) an organization dedicated to not-for-profit transparency.

- Compensation for 4M executives
  - Time span between Jan/2017 Aug/2019
  - Limited to full time -> 130K executive payments
- Governance data for 1.4M not-for-profits
  - Time span between 2010 2018
  - Limited to 2015 2017 -> 270K unique organizations

### **Data Assumptions**

	(B) Average hours per week (list any) hours for related organizations below dotted line)	box, u office or directo	ot ch unles	eck s per l a d	ition more	n of the state of	ee) Forme	(D) Reportable compensation from the organization (W-2/1099-MISC)	(E) Reportable compensation from related organizations (W-2/1099-MISC)	(F) Estimated amount of other compensation from the organization and related organizations
(1)										

- Identifying full time executives vs. part time trustees or directors (>40hrs per week & > median state income)
- Exclude executives with salaries over 1M
- Final number of executives compensated 130K



### **Primary Features**

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#### **Compensation:**

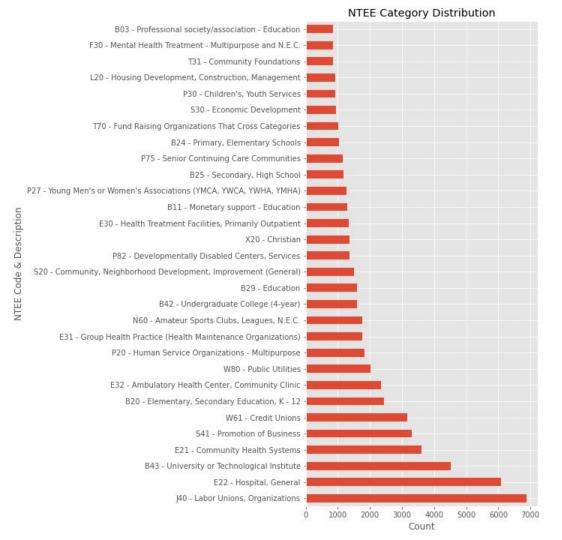
- EIN organization unique identifier
- Subsection 501(c) type
- NTEE Code categories or purpose for the organization
- State
- Formation year
- Assets
- Liabilities
- Expenses
- Revenue

#### **Governance:**

- EIN organization unique identifier
- Voting Member Count board of directors
- Employee Total Count
- Volunteer Total Count
- Salaries Expense
- Governance controls(Yes/No):
  - Conflict of interest policy?
  - o Whistleblower policy?
  - CEO Compensation review?
  - Document Retention Policy?
  - Delegation of Management Policy?
  - o etc.

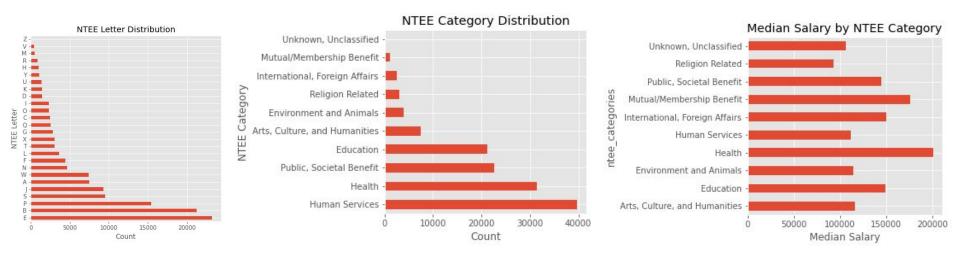
# Data Exploration - NTEE Designation

- NTEE (National Taxonomy of Exempt Entities) codes represent the mission or purpose of the not-for-profit (over 600 unique codes)
- The taxonomy was created in the 1980s.



### Feature Engineering - NTEE Designation

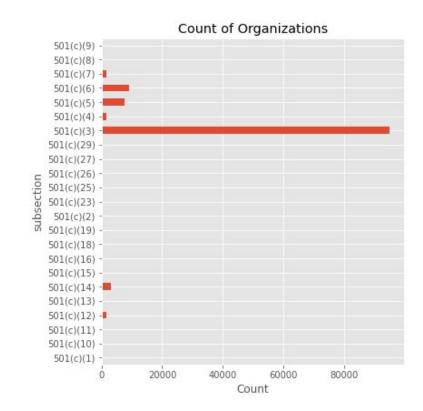
- Extracted letter designation. After modeling this level of granularity wasn't helpful.
- As a result, I wrote a function to <u>consolidate</u> the letters into 10 broad categories



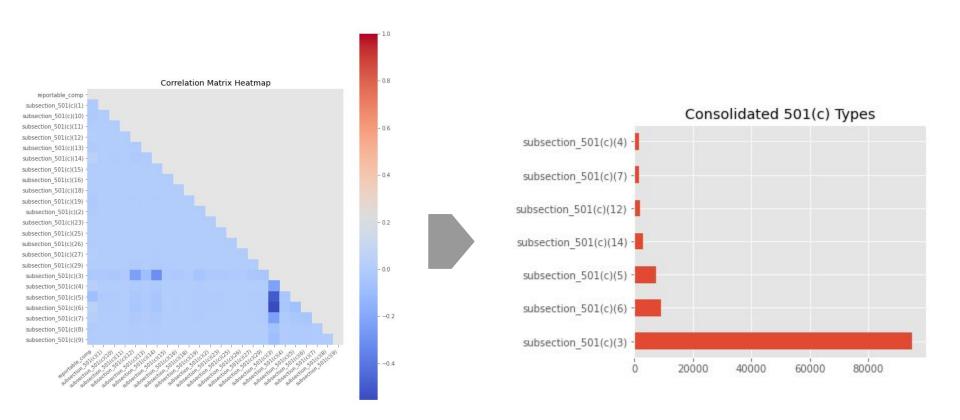
### Data Exploration - 501(c) type heavily imbalanced

Many 501(c) types can be highly specific and sometimes obscure:

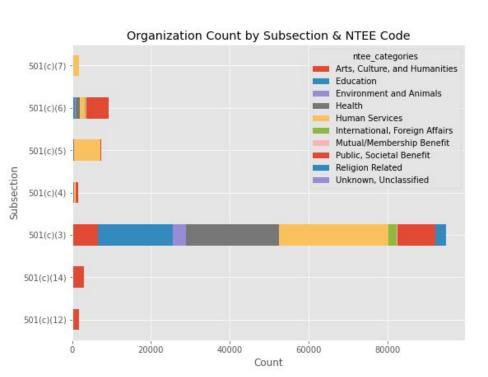
- 501(c)(6) Business Leagues, Chambers of Commerce, Real Estate Boards, etc. The NFL is included within this subsection.
- 501(c)(5) Labor and Agricultural Organizations
- 501(c)(14) Credit Unions
- 501(c)(3) religious, charitable, scientific, literary, or educational purposes, for testing for public safety, to foster national or international amateur sports competition, for the prevention of cruelty to children, women, or animals.
- 501(c)(23) Any association organized before 1880 for the purpose of providing insurance and other benefits to veterans or their dependents

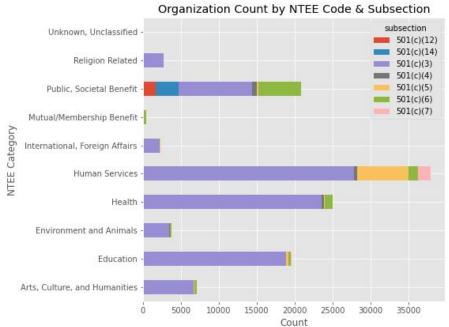


### Feature Engineering - Consolidating 501(c) Subsections



### Data Exploration - 501(c) type & NTEE Category Interaction

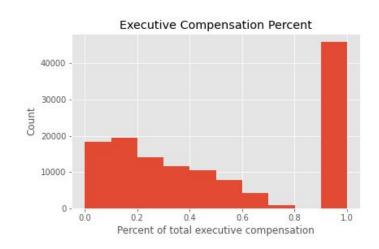


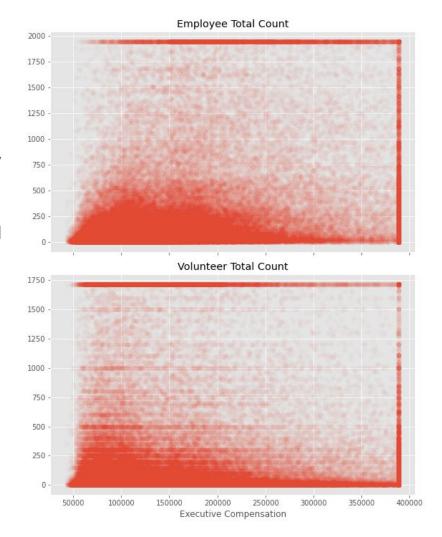


### Data Exploration: Continuous Features

Almost half of all executives are the only executive at the not-for-profit.

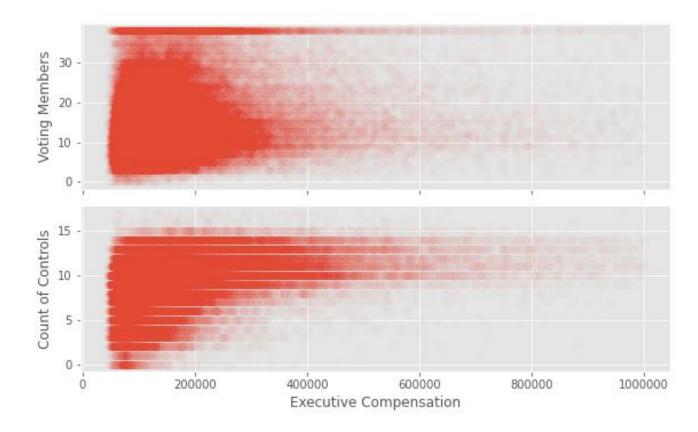
Lower executive compensation is associated with more volunteers.





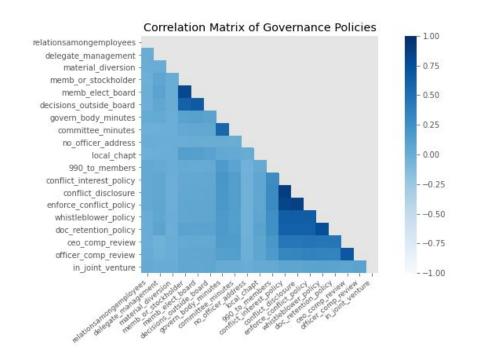
### **Data Exploration: Governance Controls**

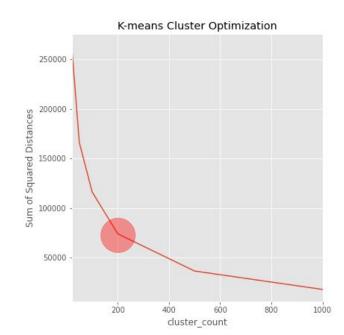
Executive compensation increases as governance controls increase.



### Feature Engineering: Governance Feature Clusters

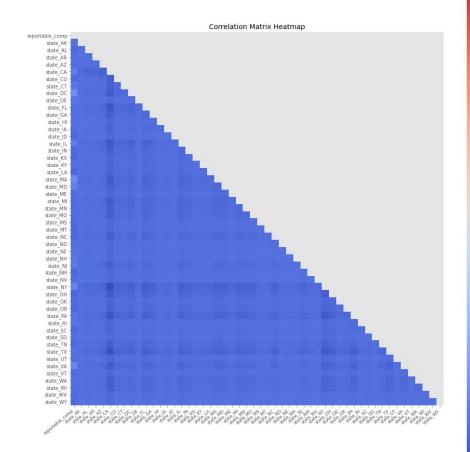
Clustering governance data into 200 clusters using K-means





## Feature Engineering: Not-for-Profit Location

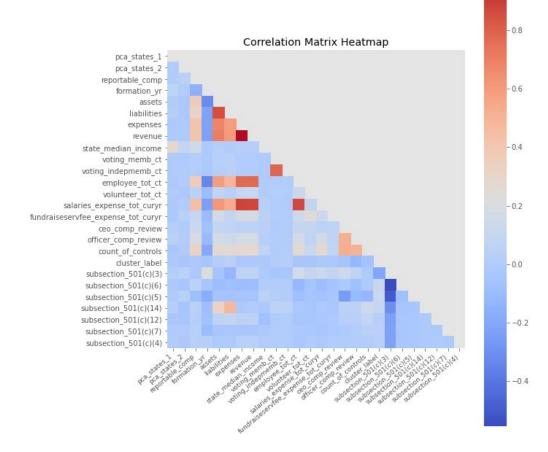
- Top correlated
   States
   consolidated into
   PCA features.
- Obtained median salaries by State dataset, joined that as a feature to add additional location information.



# Feature Engineering: Continuous Features

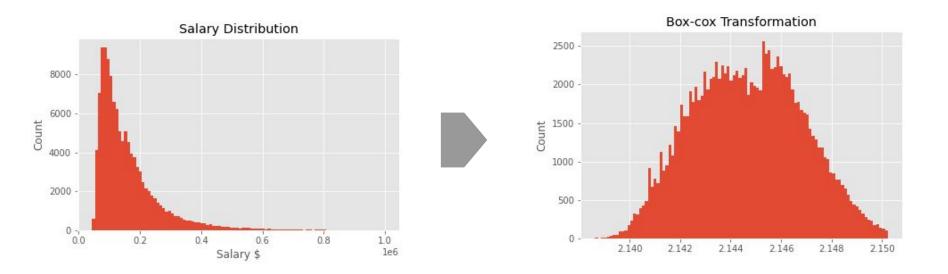
Consolidated top correlated continuous features into 3 principal components.

- Assets
- Liabilities
- Revenue
- Salaries expense
- Expenses
- Employee total count



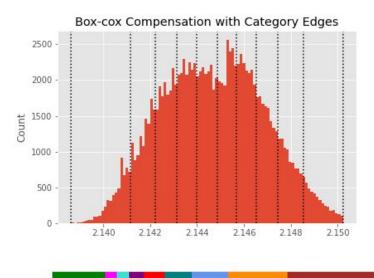
### **Target Category Levels**

What salary levels should executives be categorized into? Because the salaries aren't normally distributed, I transformed with box-cox.



### **Target Category Levels**

Found breakpoints using <u>Jenks</u> method with bins set to 10.

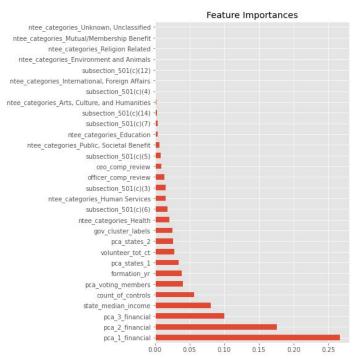


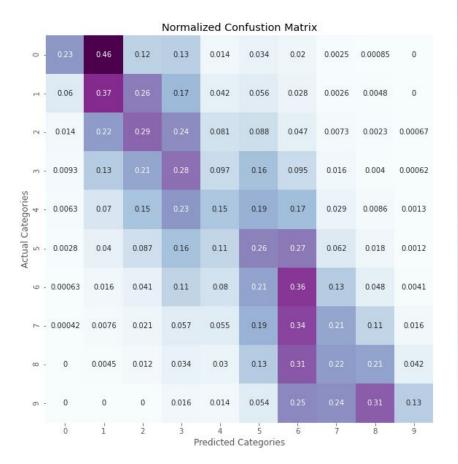
Salary Category Cut Off Points (\$)



### Model 1: Random Forest Classifier

Train, Test Score: •35, •26





- 0.4

- 0.3

-02

- 0.1

-0.0

- 0.30

-0.25

0.20

- 0.15

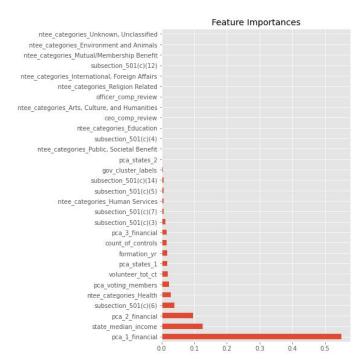
- 0.10

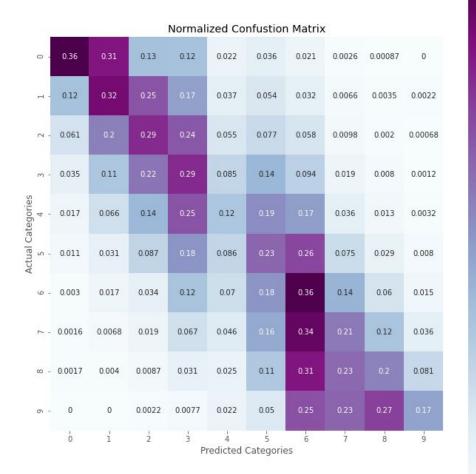
-0.05

-0.00

### Model 2: Gradient Boost Classifier

Train, Test Score: .283, .25

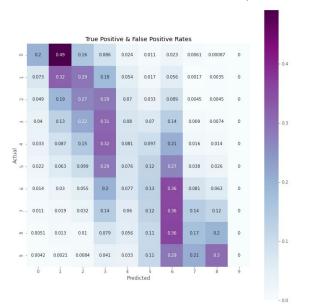




### Less Successful Models

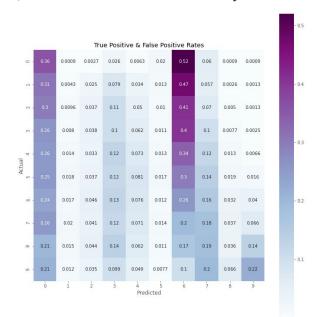
#### Logistic Classifier:

Train, Test Score: .215, .215



#### <u>Support Vector Classifier:</u>

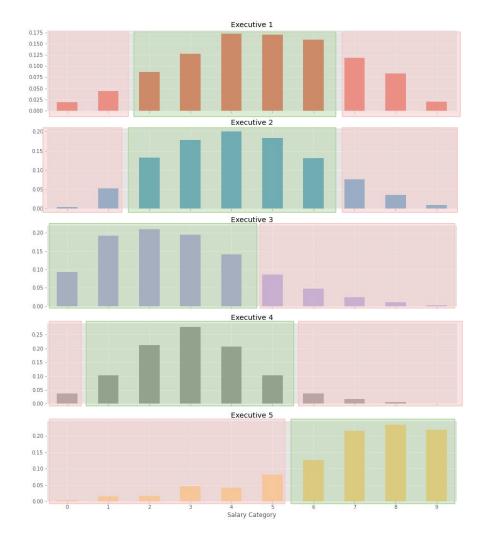
Train, Test Score: **.112**, **.109** 



# Model Application : Finding an Outlier

Examining the probability of prediction gives a sense of the confidence of the classifier.

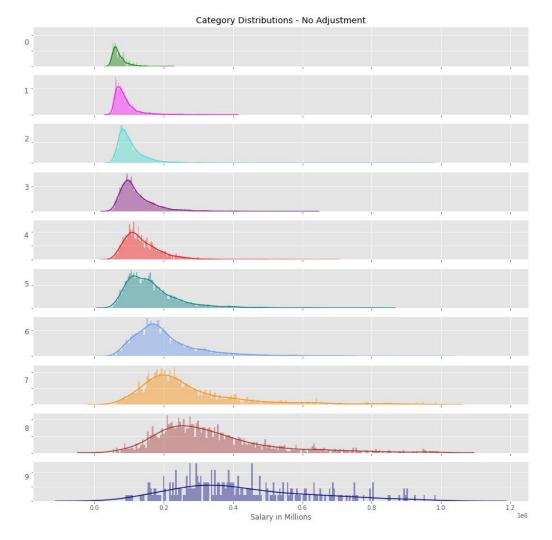
As an application for **Charity Navigator**, this could be incorporated as part of its grading process. If the actual salary falls farther than two categories from that predicted by the model, this could become a trigger for further investigation of qualitative information that might justify an outlier salary. Otherwise it might result in a hit to the not-for-profit for an unreasonably compensated or under compensated executive.



# Model Application : Salary Distributions 1

The information about not-for-profits is generalized to a range of possible values for each category. Although strong skews exist, each category has a somewhat normal distribution.

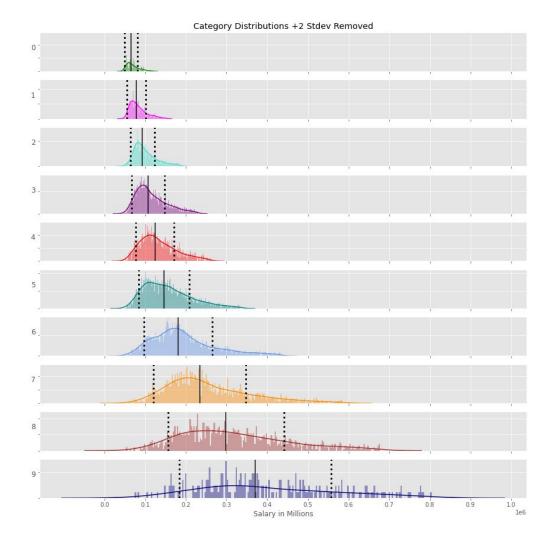
As a tool for a **Board of Directors** trying to determine what to pay an executive, the distributions can seen as ranges of possible values based on the models classifications.



# Model Application : Salary Distributions 2

To make the distributions more effective as a range of possible values:

- removed outliers above 2.5 standard deviations
- assigned median salary point for each category
- assigned standard deviation points encompassing approximately 65% of salaries for each category.



### Why Couldn't I Get Better Results?

- Motivations for involvement in a not-for-profit are often intangible.
  - a. It represents a cause they care deeply about
  - b. Ego boost is hard to quantify
- 2. Not-for-profits often raise a large share of their money through the connections of its top managers.
  - a. The value of connections can sometimes be hard to quantify.
- 3. Executive title on the tax return is currently a fill in the blank field.

  A more specific description could reduce ambiguity.
- 4. Features that I thought would be highly beneficial for classification purposes proved to be minimally effective.
  - a. NTEE category
  - b. Governance cluster labels

### **Next Steps/Other Considerations**

- Quantify not-for-profits that provide the most prestige to executives.
   Provide a survey to not-for-profit executives asking them to rank the reasons for their involvement in that particular organization.
- Obtain data to calculate how much of the not-for-profits income is derived directly from the connections of its executives.

### Questions?