# Scraping Glassdoor Job Listings

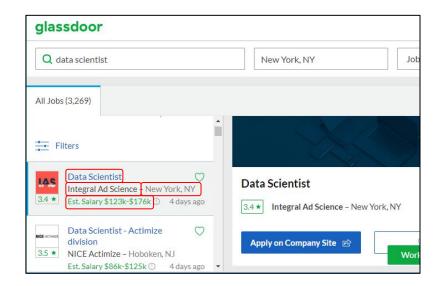
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Who wants to find a job?

John Merrick

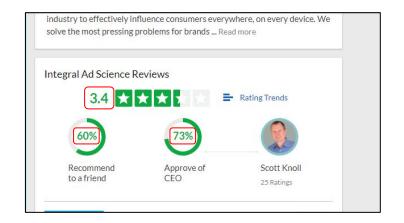
### The Plan: What to Scrape?

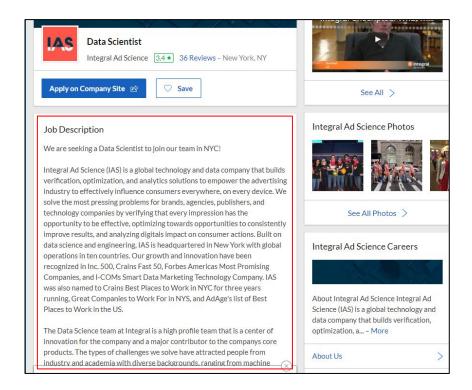
- Glassdoor job posting data using Scrapy
- Three-tier scraping algorithm:
  - Nine major metro areas
    - NYC, LA, SF, Santa Clara, Boston, Chicago, Seattle, Denver, Atlanta
  - Thirty Job Postings per search results page
  - o 15 to 35 search results pages per metro area
  - Job post details for each listing
  - Eight data points:
    - Company, title, location, est. salary, rating, CEO approval, recommend friend, and job description



### The Plan: What to Scrape?

- Returned ~7,500 job postings
- But this included ~2,500 clinical research scientist jobs





### The Plan: Cleaning the data

- Extracted high and low est. salary to calculate midpoint and range
- Create T/F columns on keywords to eliminate scientific/clinical research postings
  - Biolog, pharm, clinic, immun, chemistry, oncology, biochemistry, neuro, medical, disease, physician, surgeon, nurse, hospital, cancer, vaccine, protein, specimen
  - If a posting had more than five of these indicators, I considered the job not relevant to the analysis and dropped the row
- Create T/F columns to determine if a job requires specific skills
  - Python, scikit, R, SQL, matlab, SAS, tableau, machine learning, natural language processing, hadoop, spark, java, mongo, hive, linux, statistics, visualization, PhD, and masters
  - If a posting had fewer than two of these indicators, I considered the job not relevant to the analysis and dropped the row

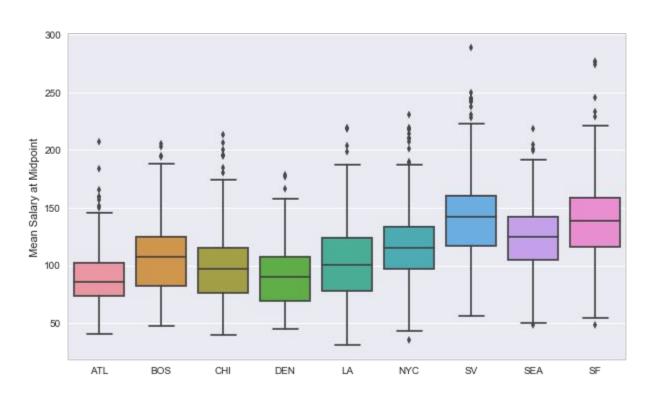
# Analysis

### **Term Frequency**

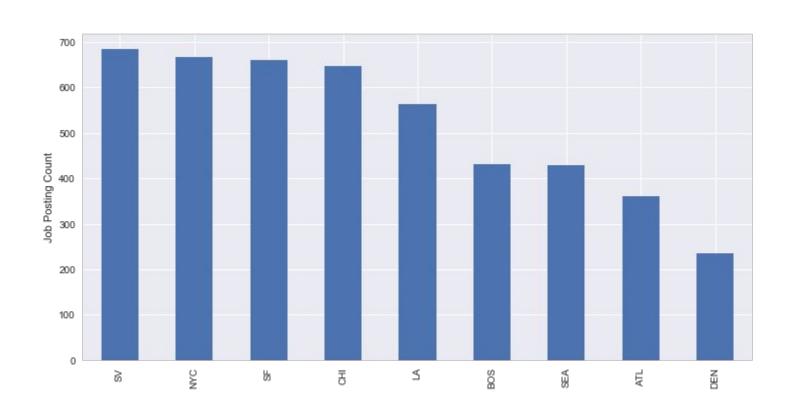
Common terms across all job descriptions

```
ance andor opportunities customers required customer modeling drive
technologies services description well result insights computerknowledge algorithms projects software product related looking
      projects software product related programming looking
industry -
                 management degreestatus
                                       develop sql
                engineering
     project complex information products
                communication understanding
```

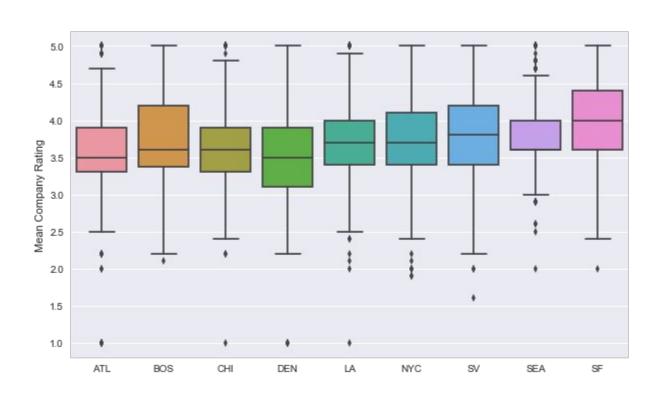
### Where to Work: Salary Distribution



### Where to Work: Posting Count



### Where to Work: Distribution of Company Ratings

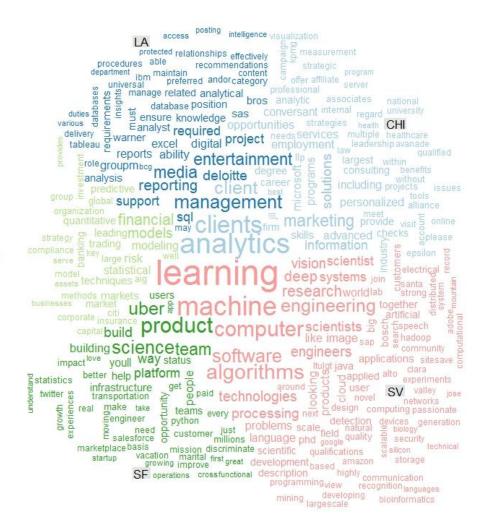


### **Region: Term Comparison**

 LA: entertainment, digital media, consulting, advertising.

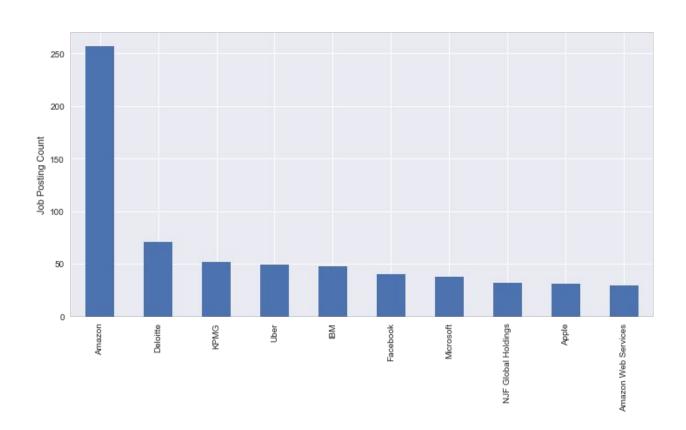
NYC

- NYC: compliance, banking, trading, risk management.
- SF: Uber, tech industry.
- SV: tech industry (machine learning, AI, NLP)
- CHI: strategy consulting, marketing,



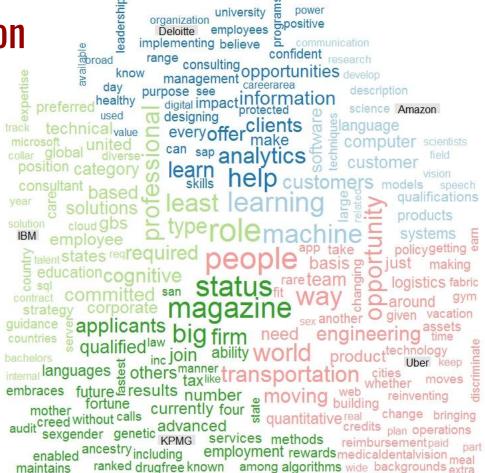
### Who is Hiring?

### Who is Hiring: Posting Count by Company



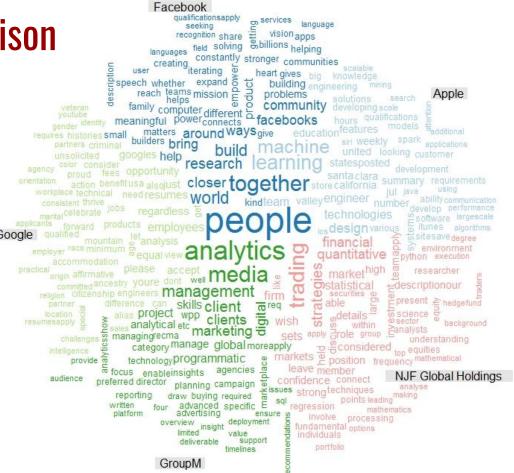
### **Company: Term Comparison**

 Each word is shown under the company that used it the most (relative frequency).



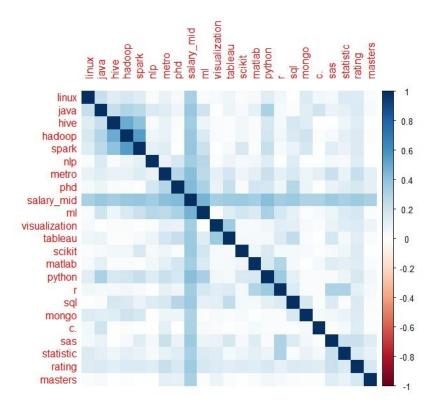
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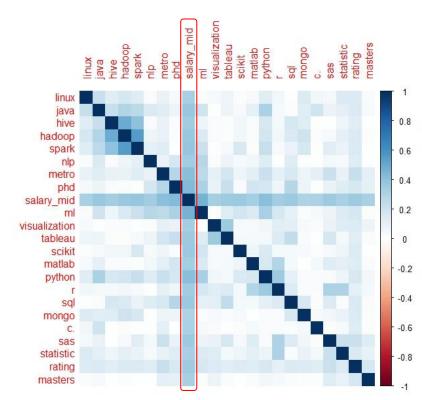
### Relationships Between the Variables

- How are the boolean "skills" columns related?
- CorrPlot suggests correlation is near zero or modestly positive for most variables.



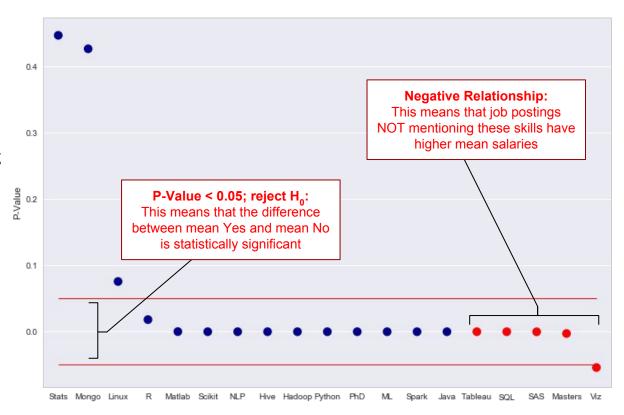
### **Relationships Between the Variables**

- How are the boolean "skills" columns related?
- CorrPlot suggests correlation is near zero or modestly positive for most variables.
- Salary appears positively correlated with all.



### Two-Sample T-Tests: Significance of P-Values

- Assumes sample SDs are equal:
  - Largest difference is NLP.
  - o 32.9 for Y, 37.0 for Y.
- Assumes population of is normally distributed.
  - Uncertain without delving deeper into the overall job market
- Assumes samples are randomly drawn and independent.
  - Also uncertain, without knowing more about how Glassdoor assigns estimated salaries to job postings.



## Conclusions

#### **Conclusions**

- Learn Python, R, Scikit, Machine Learning, NLP, Java, Spark....
  - o (and maybe get a PhD)
- SAS, Tableau, SQL, and Master's degrees tend to be associated with lower salaries.
- Salaries tend to be higher in Silicon Valley, San Francisco, and Seattle.

### **Conclusions**

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- SAS, Tableau, SQL, and Master's degrees tend to be associated with lower salaries.
- Salaries tend to be higher in Silicon Valley, San Francisco, and Seattle.
- Opportunities for future work:
  - Multiple linear regression: test predictive power of variables on salary.
  - Sentiment analysis: search for differences in tone of job description text by location, title, or company.
  - Topic modeling: use TD/IDF and cosine similarity to do a more detailed analysis on the similarities and differences between postings.
  - Recommendation engine: allow users to select skills, and return suitable job postings.