Indeed Web Scraper: Job Market Trends for Data Careers

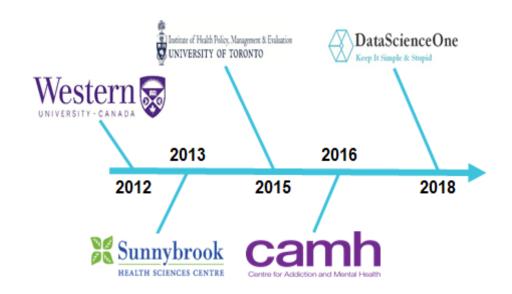
Taesun Yoo

- Aug 30, 2018 -



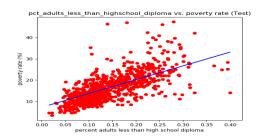
About Myself: Taesun Yoo

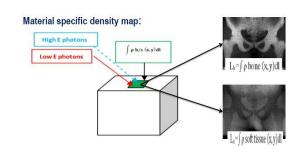
- Former BI QA Analyst, ML Enthusiast
- Founder of DataScienceOne (Youtube Channel)
- Completed Master's in Health Informatics
- Research experience: Sunnybrook
 - Medical Imaging (image processing)
 - □ Radiation Physics (cancer treatment)
- Work experience: CAMH
 - Business Intelligence
 - QA data warehousing
 - Data visualization/reporting



Kicking some side machine learning projects ...

MajorityVote Classifier		
	Predicted Class	
Actual Class	Stroke	Non-stroke
Stroke	41%	9%
Non-stroke	11%	39%







Agenda

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Problem Overview

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• Data Wrangling: Indeed.com

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Exploratory Data Analysis

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Future Work and Recommendations



Problem Overview



Indeed Job Search Results in Toronto

On Aug 26, 2018

95 jobs Data Scientist



90 jobs Data Engineer



136 jobs Data Analyst



647 jobs Business Intelligence





Indeed Job Market: Trend Analysis

Why should we care?

- Improve % of getting an interview
- Strategize job application submissions

Stakeholders:

- Job applicants
- Hiring managers and recruiters

Goal:

Provide overall data career market insights in Toronto!

Objectives:

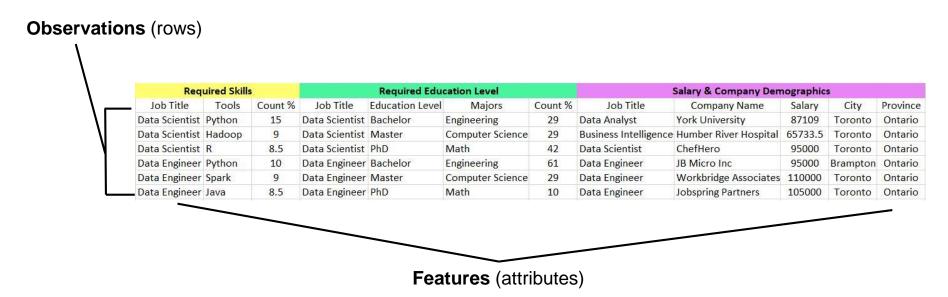
- Save time job application submission vs. keyword-based search
- Prioritize and learn top required skills/tools



Indeed Web Scraped: Dataset Overview

Dataset contains 10 variables for exploratory data analysis:

- 7 categorical & 3 numerical features
- Demographics: required education, tools/skills, salary and company
- Sample size: x < 1,000 jobs



Challenges:

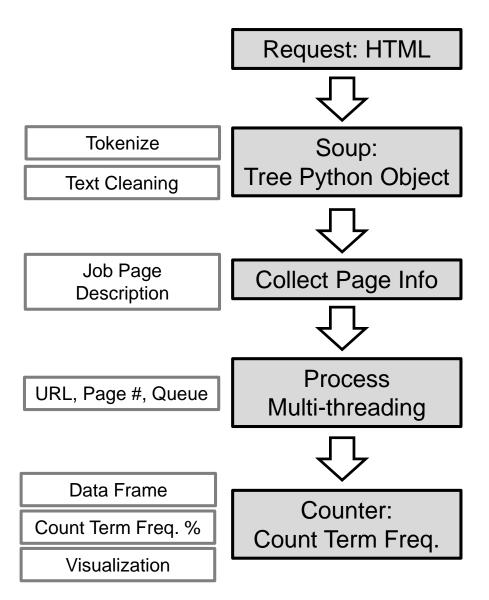
- Missing salary information (job posting)
- Inconsistent format (i.e., hourly/weekly salary)
- Duplication (i.e., same job postings)



Data Wrangling: Indeed.com



Data Wrangling Pipeline





Term Freq. =
$$\frac{Term}{\sum_{1}^{n} Term}$$
 x 100%



Sample Code: Getting Page Info

Input Parameters

```
# Function 7: iob scraper and visualize data #
def indeed_job_scraper city=None, state=None, job title=None):
    city copy = city[:]
   if city is not None:
       # For city name like 'San Francisco', we want to convert it into 'San+Francisco'
        city list = city.split()
        city = '+'.join(city list)
        url list = ['http://www.indeed.ca/jobs?q=', job title, '&l=', city, '%2C+', state]
        url list = ['http://www.indeed.ca/jobs?q=', job title]
    url = ''.join(url list) # URL for job search
    print("Using URL " + url)
        html = requests.get(url).text
    except:
        print('The location ' + city copy + ', ' + state + ' could not be found.')
       return
    soup = make soup(html)
    num = soup.find(id = 'searchCount')
    if num == None:
        num = soup.find(id = 'searchCount')
    num jobs = num.string.encode('utf-8')
    # Total number of jobs found
    job numbers = re.findall(' \setminus d+', str(num jobs))
    # Process commas in large number representations
   if len(job numbers) > 3:
        total num jobs = (int(job numbers[2]) * 1000) + int(job numbers[3])
        total num jobs = int(job numbers[2])
    if city is None:
        print(str(total_num_jobs) + ' jobs found nationwide')
    print(str(total_num_jobs) + ' jobs found in ' + city_copy + ', ' + state)
    num_pages = int(total_num_jobs / 10)
   # Obtain company list
    page company = []
    for element in soup.find_all('span', class_='company'):
        page_company.append(element.get_text().strip()) # append company list
```

```
# Multi-threading #
total job descriptions = process url(num pages, url) # Four-dimensional list
# Convert four-dimensional list into two-dimensional list
total job descriptions = sum(sum(total job descriptions, []), [])
total jobs found = len(total job descriptions)
print('Done with collecting the job ads!')
print('There were ' + str(total jobs found) + ' jobs successfully found.')
         Term Frequency Counter
doc freg = Counter()
[doc freq.update(item) for item in total job descriptions]
## List of company names: numjob openings ##
comp dict = Counter(page company)
# Convert results into a dataframe
df comp = pd.DataFrame.from dict(comp dict, orient='index').reset index()
df comp.columns = ['CompName', 'NumJobs']
# sort data for plotting
df_comp.sort_values(by = 'NumJobs', ascending=False, inplace=True)
pd.set option('display.width', 1000)
                     Output #1: Education Level
## Requirement by education level: calculate % apperance in job ads ##
educ dict = Counter({'PhD': doc freg['phd']+doc freg['doctor']+doc freg['doctorate'],
                    'Master': doc_freq['msc']+doc_freq['master'],
                    'Bachelor': doc freg['bsc']+doc freg['bachelor']+doc freg['university']}
df educ = pd.DataFrame(list(educ dict.items()), columns = ['Educ Level', 'NumJobs'])
# Percentage of job ads having a education requirement
df_educ['Percentage'] = (df_educ.NumJobs / df_educ.NumJobs.sum()) * 100
# sort data for plotting
df_educ.sort_values(by='Percentage', ascending=False, inplace=True)
pd.set_option('display.width', 1000)
```

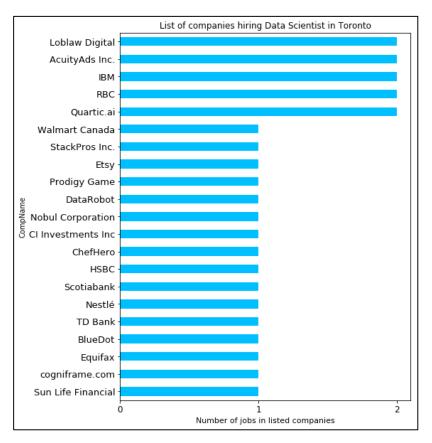


Exploratory Data Analysis

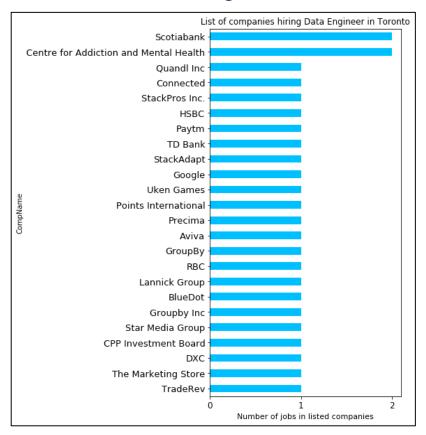


Hiring Companies: Data Scientist | Engineer

Data Scientist



Data Engineer



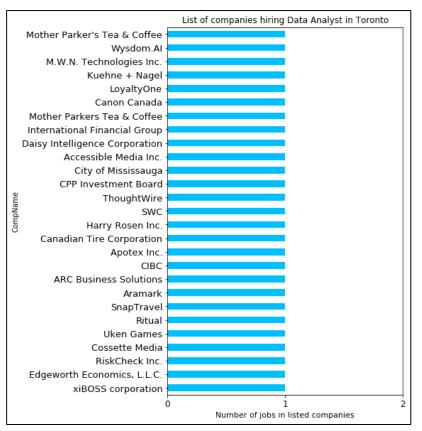
Companies hiring ML talents in Toronto

- Retail: Loblaw, Walmart, ...
- Finance: TD Bank, HSBC, ...
- Insurance: Sun Life, Aviva
- Start-Ups: LoyaltyOne, Precima
- Healthcare: CAMH

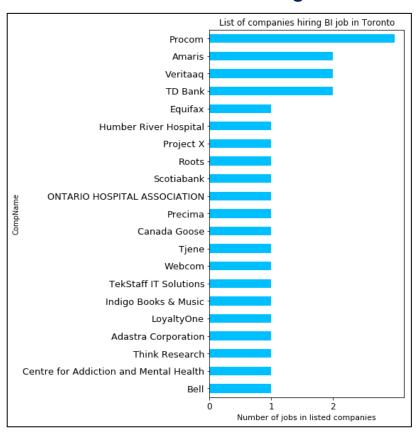


Hiring Companies: Data Analyst | Bl Job

Data Analyst



Business Intelligence



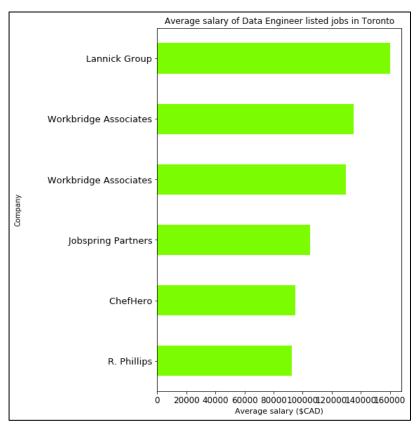
Companies hiring Bl/analytics talents in Toronto

- · Retail: Canada Goose, Harry Rosen, ...
- Finance: TD Bank, HSBC, ...
- Consulting Firm: Adastra, ...
- Tech: Amazon, ...
- Government: City of Mississauga, ...



Salary: Data Scientist | Engineer

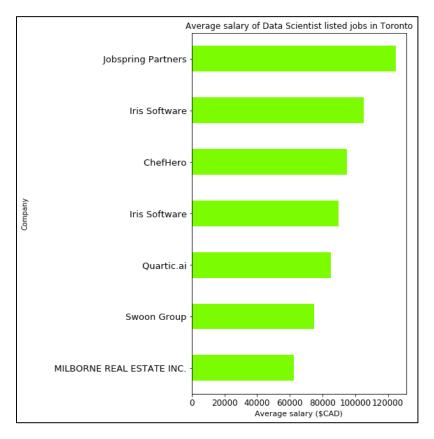
Data Engineer



Average salary

\$ 119,500 per year

Data Scientist



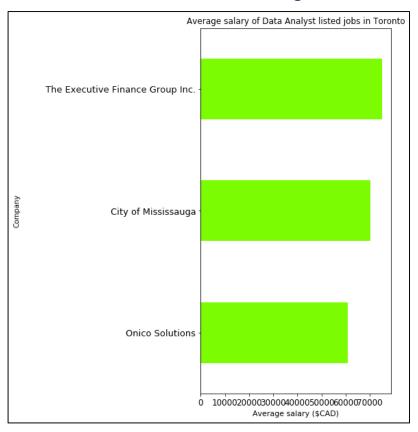
Average salary

\$91,071 per year



Salary: Data Analyst | BI Jobs

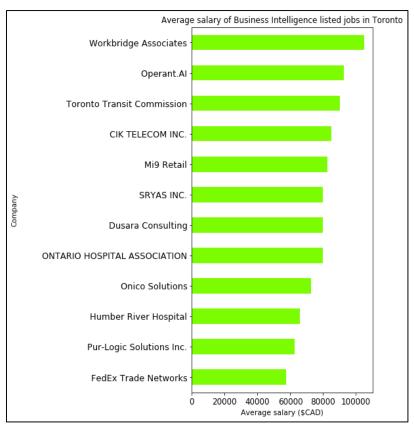
Business Intelligence



Average salary

\$79,426 per year

Data Analyst



Average salary

\$68,722 per year



Skill Requirement: Data Scientist | Engineer



70

80

60

50

Percentage Appearing in Job Ads (%)

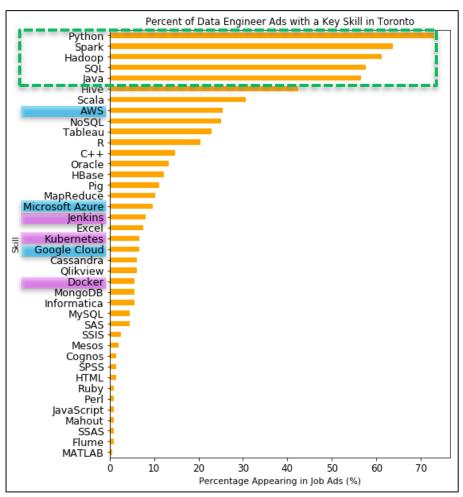
Data Scientist

Percent of Data Scientist Ads with a Key Skill in Toronto Python SQL Hadoop Spark java SAS. Hive Scala Tableau Microsoft Azure Oracle Excel AWS HBase Jenkins MongoDB MATLAB HTML Cognos C++ SPSS MapReduce NoSQL Google Cloud Qlikview JavaScript Ruby Pig MySQL SSAS SSIS PHP STATA SAP

20

10

Data Engineer





Docker

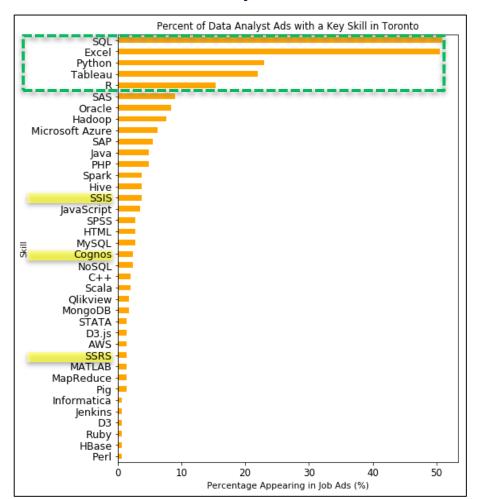
Kubernetes

Cassandra

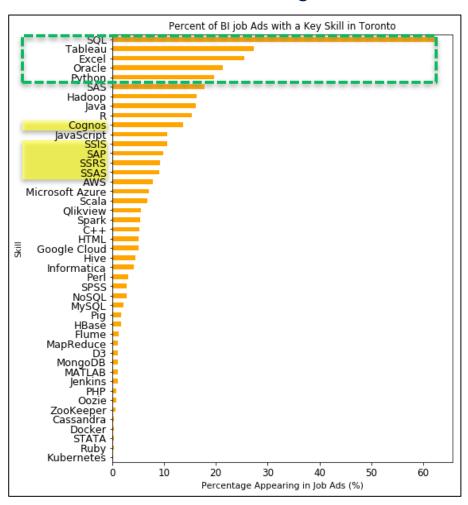
Skill Requirement: Data Analyst | Bl Jobs



Data Analyst



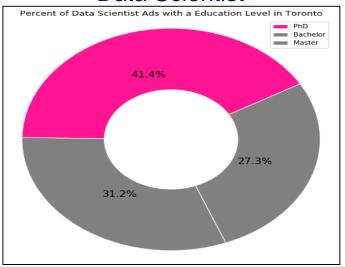
Business Intelligence



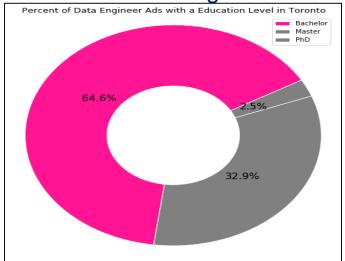


Education Requirement: Levels

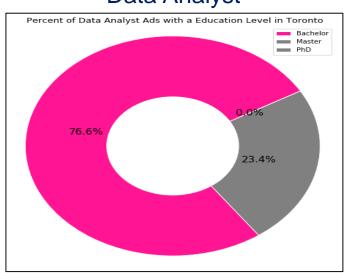
Data Scientist



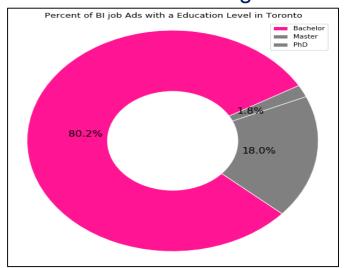
Data Engineer



Data Analyst



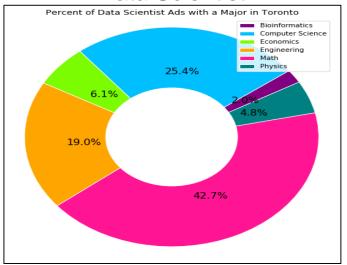
Business Intelligence



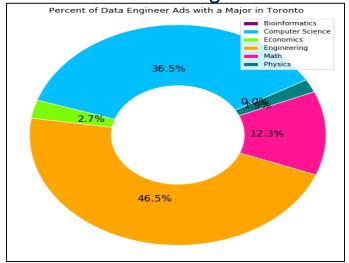


Education Requirement: Majors

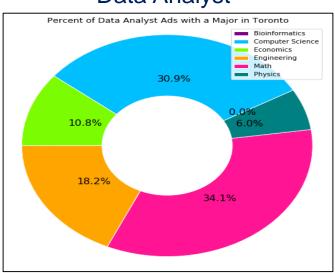
Data Scientist



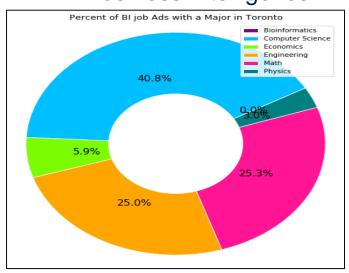
Data Engineer



Data Analyst



Business Intelligence





Summary: Indeed Job Scraper

Goal

Provide overall data career market insights in Toronto

Key Insights

- Most of sectors hiring for ML/BI talents are from finance, insurance, retail, etc.
- Avg. annual salary: 1st rank data engineer (\$119K) vs. 4th rank data analyst (\$69K).
- Required skill set:
 - Data engineer focused more on DevOps and cloud experience
 - ➤ BI job focused more on ETL tools (i.e., Oracle, Microsoft BI stack)
- Required education level & majors:
 - Data scientist was only job with preferred qualification at PhD level
 - Most of data career jobs preferred majors were from S.T.E.M field



Future Work & Recommendations



Limitations & Future Work

Limitation:

- Limited # of job ads with salary info.
- Sample size is not at large scale.

Future Work:

- Data sources: consider web scraping other sites such as
 - MonsterJob, GlassDoor and LinkedinJob
- Data collection:
 - Company's review comments by employees
 - Company's roles and environment ratings by employees
- ML modeling:
 - Sentiment analysis on positive and negative comments
 - LDA topic modeling and word cloud by job categories
 - Data career salary regression model by job categories



Recommendations

Strategic job application submissions

- Understand education requirement before applying
- Learn required tools by a job (ranked importance)
- Compare average salary of your targeted companies

Design ML model portfolios

- Salary regression model
- Sentiment analysis on review comments
- Cluster analysis by job categories

Data collection

- Consider other data sources: Linkedin, Glassdoor, etc.
- Collection of data on company's review comments/ratings



Next Episode

Boston Housing Price Prediction Model with Regression

