

Data Science in Industry

Software Engineer Salaries in Washington, DC Area

6,922 Salaries Updated Dec 4, 2018

Industries

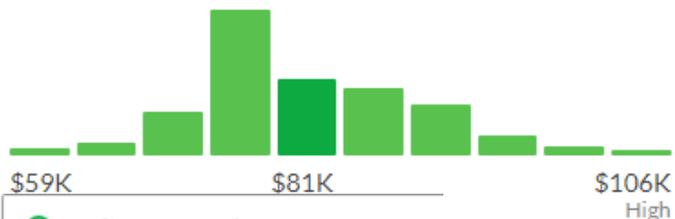
Company Sizes

0-1 Years

Average Base Pay

\$81,360 /yr

7% below national average



software engineer

Additional Cash Compensation

Average \$7,049

Range \$1,618 - \$18,353

Job Type

Full-time (12061)

Part-time (89)

Contract (36)

Internship (112)

Temporary (8)

Apprentice/Trainee (19)

Entry Level (100)

Data Scientist Salaries in Washington, DC Area

210 Salaries Updated Oct 3, 2018

Industries

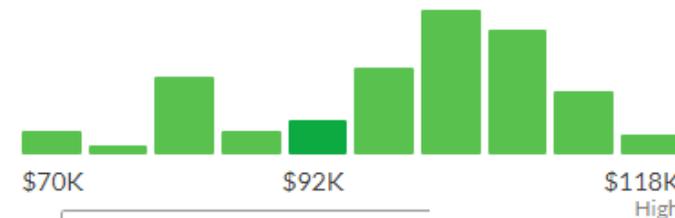
Company Sizes

0-1 Years

Average Base Pay

\$92,158 /yr

12% below national average



data scientist

Additional Cash Compensation

Average \$9,627

Range \$3,195 - \$22,206

Job Type

Full-time (2084)

Part-time (13)

Contract (2)

Internship (17)

Temporary (1)

Apprentice/Trainee

Entry Level (10)

Keywords

data scientist

Open to the public ✕ 20740 ✕ 25 miles ✕

✕ Remove all filters

Viewing 1 - 10 of 16 jobs

Save this search. We'll email you new jobs as they become available.

Data Scientist

Central Intelligence Agency

Other Agencies and Independent Organizations

Washington DC, District of Columbia

Open 03/15/2018 to 03/14/2019

Data Scientist

National Geospatial-Intelligence Agency

Department of Defense

Multiple Locations

Open 10/30/2018 to 12/29/2018

GENERAL PHYSICAL SCIENTIST

U.S. Navy - Agency Wide

Department of the Navy

Multiple Locations

Open 11/02/2018 to 11/01/2019

INTERIOR DESIGNER

U.S. Navy - Agency Wide

Department of the Navy

Multiple Locations

Keywords

software engineer

Open to the public ✕ 20740 ✕ 25 miles ✕

✕ Remove all filters

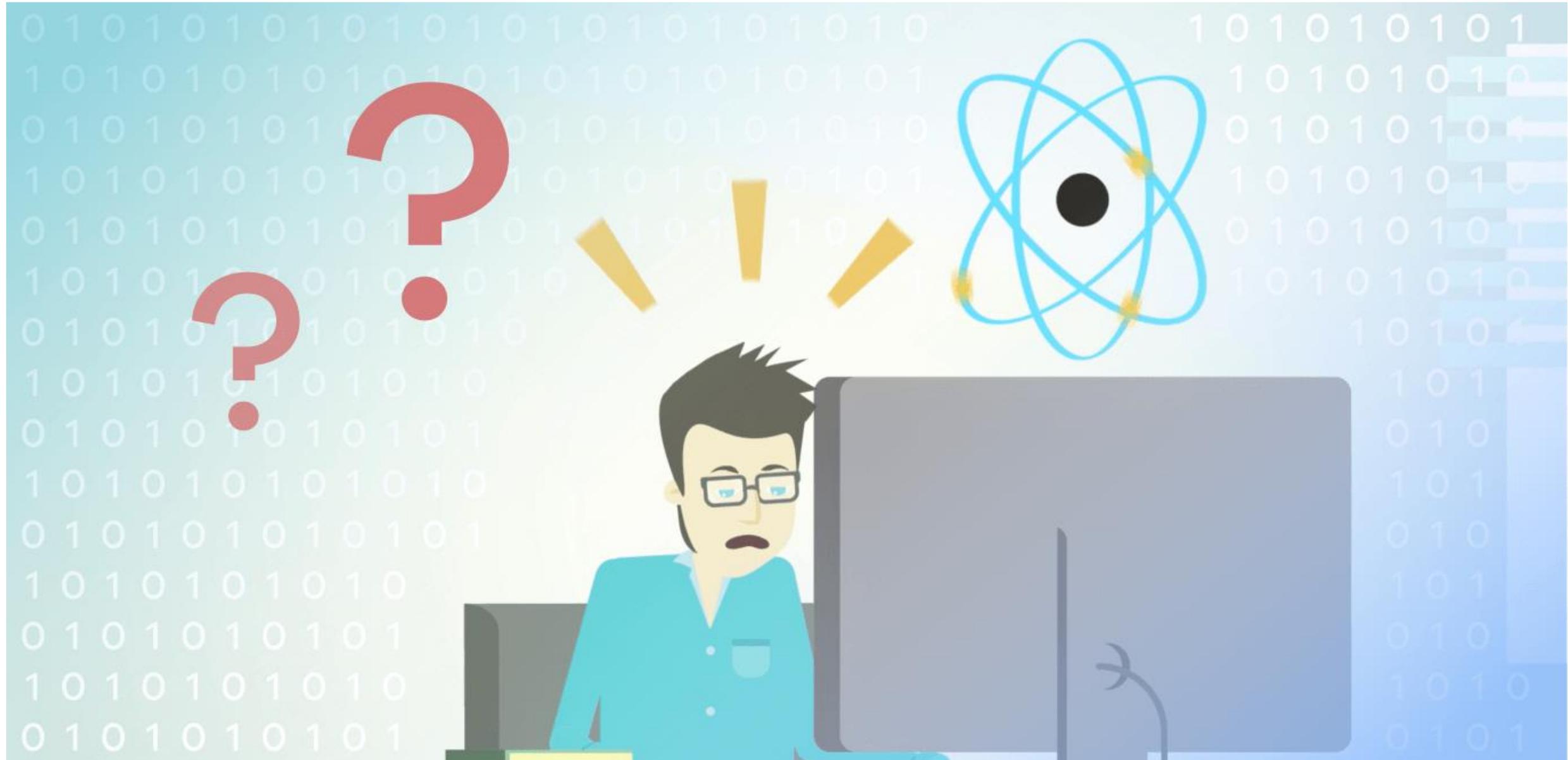
Viewing 1 - 10 of 13 jobs

Filter results

CATEGORY

- Data Science (80)
- Design (18)
- Engineering (567)
- Finance (61)
- Legal (15)
- Operations (21)
- People & Workplace (116)
- Personal Banking (27)
- Product Management (136)
- Project/Process Management (121)
- Risk Management & Audit (103)
- Software Engineering (116)

What is the hardest part of data science?



What is the hardest part of data science?

1. Dealing with people 😊

2. Figuring out what questions to ask (domain knowledge)
3. Getting the data for those questions
4. Organizing the data
5. Dealing with missing data
6. Training supervised Machine Learning

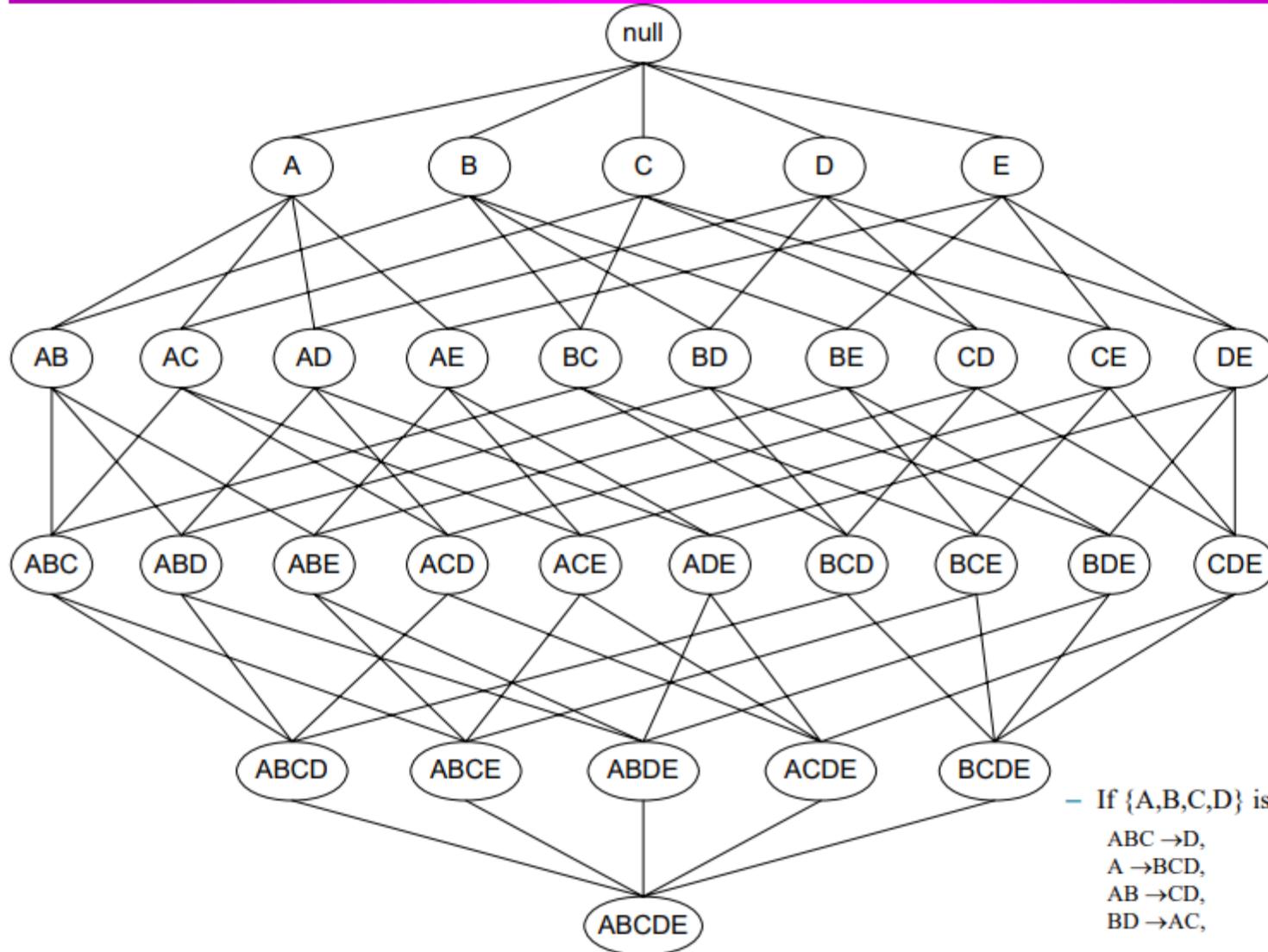
How to become successful?



Association Rules

1. Gather all frequent item sets (specified by support % of occurrences)
2. From those frequent items consider each possible combination and calculate the supports.
3. The most frequent rules that match our support level are presented.

Basic idea behind rule generation



- If $\{A,B,C,D\}$ is a frequent itemset, candidate rules:
 - $ABC \rightarrow D,$ $ABD \rightarrow C,$ $ACD \rightarrow B,$ $BCD \rightarrow A,$
 - $A \rightarrow BCD,$ $B \rightarrow ACD,$ $C \rightarrow ABD,$ $D \rightarrow ABC$
 - $AB \rightarrow CD,$ $AC \rightarrow BD,$ $AD \rightarrow BC,$ $BC \rightarrow AD,$
 - $BD \rightarrow AC,$ $CD \rightarrow AB,$

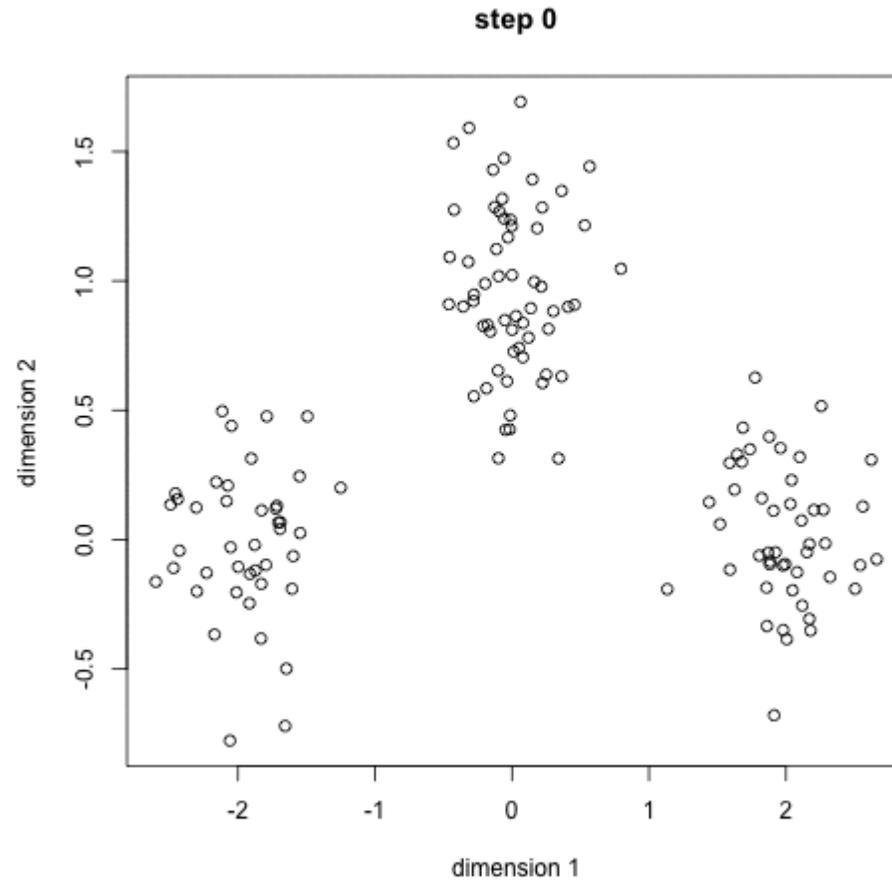
Association rule terminology

- Confidence: $\text{Confidence}(x \rightarrow y) = \text{support}(x \cup y) / \text{support}(x)$
- Lift: $\text{Lift}(x \rightarrow y) = \text{support}(x \cup y) / (\text{support}(x) * \text{support}(y))$
 - If the rule had a lift of 1, it would imply that the probability of occurrence of the antecedent and that of the consequent are independent of each other. When two events are independent of each other, no rule can be drawn involving those two events.
- Conviction: $\text{conv}(X \Rightarrow Y) = \frac{1 - \text{supp}(Y)}{1 - \text{conf}(X \Rightarrow Y)}$
 - interpreted as the ratio of the expected frequency that X occurs
 - without Y (that is to say, the frequency that the rule makes an incorrect prediction)
 - if X and Y were independent divided by the observed frequency of incorrect
 - predictions.

Jupyter Example

Clustering

K-Means will give you clusters that you can label!



Putting it together

1. From clusters you can label them which allows you to engineer statistically the association rule role ups.
2. From there you can see all the rules and target a dependent variable
3. Set your decision tree, random forests, or neural network to target this dependent variable and independent variables
4. Update with feedback from the field and you don't have to worry about changing any code. Plus what code/variables would you change? The time savings of ML 😊.

Questions 😊