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Project- Data Scientist Salaries
IST 707 Tuesdays

Introduction:

An analysis was completed on a data set from Kaggle of Data Science salaries offered from 2019-2022. This analysis will help to predict what salaries may be offered in 2023 and what a persona may expect when negotiating for a data scientist role. Multiple analysis methods will be used to help answer the question presented on the data. This is a replacement set of data as the first data set was unable to be cleaned in a timely manner to proceed with the analysis due to text issues. The new data set was precleaned and only required a small amount of cleaning to prepare the data for different methods.

Analysis and Models

About the data

This data set includes five columns: Company, Role, Salary, City and Startdate and has 13,321 rows of data. This was already precleaned and had no empty or null data, and the salaries were free of and symbols and punctuation which all allowed for easier processing of the data.

To start the preparation of this data it was read into R and stored as a data set named ds_salary. The data was then viewed to see what was listed in the data set to do a double check to make sure there is no missing or incorrect data.

3					
^	company	role	salary [‡]	city	startdate [‡]
1	OPEN DATA GROUP INC	DATA SCIENTIST	51860	CHICAGO, IL	2019
2	BLINKAI TECHNOLOGIES INC	DATA SCIENTIST	59340	BOSTON, MA	2019
3	DSFEDERAL INC	DATA SCIENTIST	60420	ROCKVILLE, MD	2019
4	DSFEDERAL INC	DATA SCIENTIST	60420	ROCKVILLE, MD	2019
5	CYBERXDATA LLC	DATA SCIENTIST	62000	NEWTON, MA	2019
6	ADECCO GROUP NA/MODIS INC	DATA SCIENTIST	65500	DEARBORN, MI	2019

An initial review of the data was completed to see where most of the information was within the data. This was done by viewing the mean, median and Freq of the salary ranges.

Once that was completed the min, max and quartiles were viewed, and the summary of the data was displayed:

```
> sd(ds_salary$salary) #
                              standard
                                                    deviation
[1] 36440.64
> max(ds_salary$salary)
[1] 1250000
> min(ds_salary$salary)
[1] 450
                                           - min(ds_salary$salary)
> range <- max(ds_salary$salary)</pre>
> range
[1] 1249550
        <- quantile(ds_salary$salary,</pre>
> qt
                                           na.rm=TRUE)
alues
> qt
     0%
             25%
                     50%
                              75%
                                      100%
          91021 114000 142095 1250000
    450
> summary(ds_salary)
  company
                     role
                                      salary
                                                      city
                                                                      startdate
Length:13321
                 Length:13321
                                  Min. :
                                             450 Length:13321
                                                                    Min.
                                                                          :2019
Class :character Class :character
                                   1st Qu.: 91021
                                                   Class :character
                                                                    1st Qu.:2020
Mode :character Mode :character
                                   Median : 114000
                                                   Mode :character
                                                                    Median:2020
                                        : 118226
                                                                    Mean
                                                                         :2020
                                   Mean
                                   3rd Qu.: 142095
                                                                    3rd Qu.:2021
                                        :1250000
                                   Max.
                                                                    Max.
                                                                           :2022
> |
```

Initial Results before further Analysis

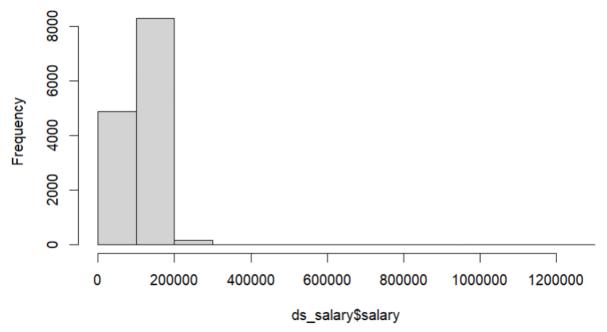
In the initial review the average salary for all the years combined was \$118,225.00. The most frequent salary was \$120,000, with 435 occurrences, and falls in the middle of the 50th and 75th percentile in the dataset. The maximum salary was \$1,250,000.

Models

The following models were created from the stored data sets to help understand the data and better analyze the data for results.

Histogram- Salary compared to frequency.

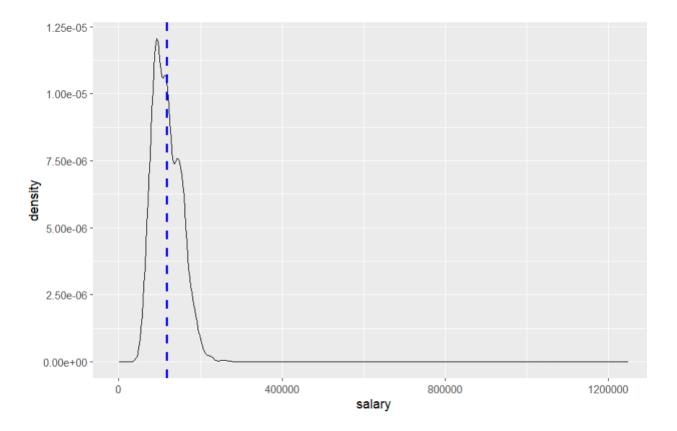
Histogram of ds_salary\$salary

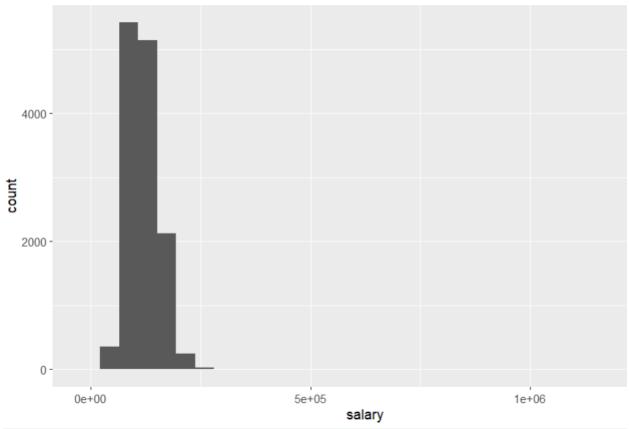


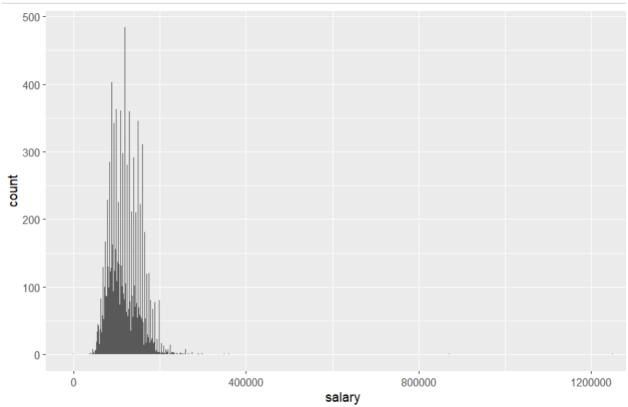
Word Cloud- Views the most frequent words (limited to 50)

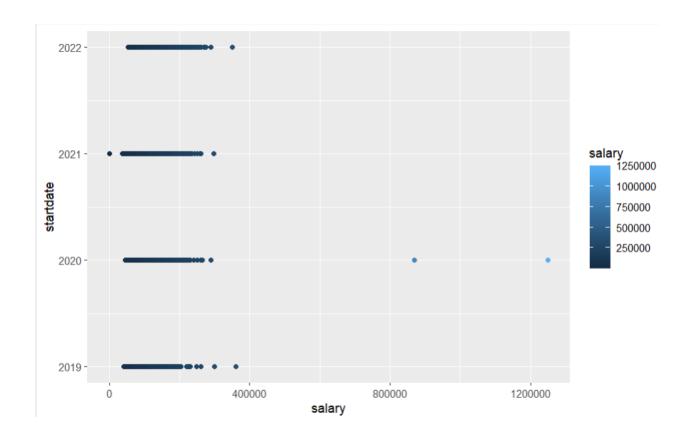


GGPLOTS-density with mean line, bar chart showing the count of the salaries from the entire dataset. Salary broken down by year, the coloring is the salary range.









Additional Models- Mean salaries per year, Mean salaries broken down by city.

```
startdate mean_salary
         \langle db 1 \rangle
                          \langle db 1 \rangle
          <u>2</u>019
                      <u>110</u>556.
2
          <u>2</u>020
                      <u>115</u>906.
3
          2021
                       <u>121</u>755.
4
          <u>2</u>022
                      <u>129</u>672.
# A tibble: 901 \times 2
    city
                                          mean_salary
    <chr>>
                                                   \langle db 1 \rangle
                                                <u>186</u>898
 1 (AKA 1601 WILLOW ROAD), CA
 2 12TH FLOOR, NY
                                                130000
 3 805 MOBERLY LANE, AR
                                                <u>104</u>000
 4 ACTON, MA
                                                <u>121</u>436
                                                 <u>94</u>069.
 5 ADDISON, TX
 6 ALAMEDA, CA
                                                <u>110</u>000
                                                 95871.
 7 ALBANY, NY
 8 ALBUQUERQUE, NM
                                                 <u>59</u>155
                                                 <u>79</u>019
 9 ALDIE, VA
10 ALEXANDRIA, VA
                                                110149.
# i 891 more rows
# i Use `print(n = ...)` to see more rows
```

Data Mining Models-Probability table, Naïve Bayes Model, Predicted Salary with Naïve bayes test data, Decision tree, and randomForest.

Probability-

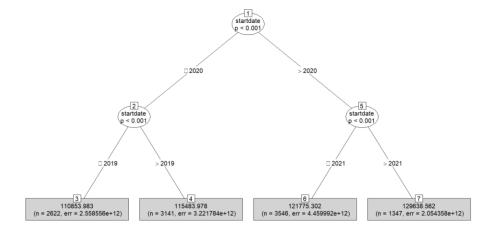
450	37149	9 40000	40706	41184	44637	45490
9.384384e-05	9.384384e-05	5 9.384384e-05	9.384384e-05	9.384384e-05	4.692192e-04	9.384384e-05
46925	48152	2 49192	50000	50627	51000	51100
9.384384e-05	2.815315e-04	4 9.384384e-05	9.384384e-05	9.384384e-05	9.384384e-05	1.876877e-04
51355	52000	53000	53539	53560	53726	53934
9.384384e-05	4.692192e-04	4 9.384384e-05	9.384384e-05	9.384384e-05	1.876877e-04	9.384384e-05
54163	54300	54392	54413	54434	54600	54787
1.876877e-04	9.384384e-05	5 9.384384e-05	5.630631e-04	9.384384e-05	1.876877e-04	9.384384e-05
54808	54850	54891	55000	55037	55120	55578
9.384384e-05	3.753754e-04	4 1.219970e-03	6.569069e-04	9.384384e-05	9.384384e-05	9.384384e-05
55723	55869	9 56000	56003	56493	56534	56880
9.384384e-05	9.384384e-05	5 9.384384e-05	9.384384e-05	9.384384e-05	9.384384e-05	9.384384e-05
57013	57096	57105	57179	57242	57450	57460
9.384384e-05	9.384384e-05	5 9.384384e-05	3.753754e-04	9.384384e-05	9.384384e-05	9.384384e-05
57741	57750	57762	57782	57977	58000	58094
9.384384e-05	9.384384e-05	5 1.876877e-04	9.384384e-05	9.384384e-05	9.384384e-05	9.384384e-05
58178	58198	8 58510	58552	58635	58826	59000
0.007000 00	0.004004 05	070077 04	4 070077 04	4 070077 04	0.004004 05	4 070077 04

			נפכנט	2020.000	0.0000000
			65416	2020.000	0.0000000
			65478	2020.000	NA
Conditio	nal probak	nilities:	65499	2019.500	0.7071068
	startdate	orrieres.	65500	2020.000	1.0000000
Υ	[,1]	[,2]	65562	2020.000	NA
450	2021.000	NA	65666	2021.333	0.5773503
37149	2021.000	NA	65811	2021.000	0.0000000
40000	2019.000	NA	65832	2019.000	NA
40706	2019.000	NA	65874	2020.000	NA
41184	2021.000	NA	66000	2019.429	0.7867958
44637		0.0000000	66040	2020.000	NA
45490	2021.000	NA	66186	2021.000	0.0000000
46925	2020.000	NA	66227	2021.000	NA
48152		0.0000000	66477	2020.000	NA
49192	2020.000	NA	66500	2019.000	NA

Prediction-

		role	salary city	startdate pred	
	<chr></chr>	<chr></chr>	<db1> <chr></chr></db1>	<db1> <fct></fct></db1>	<fct></fct>
1	OPEN DATA GROUP INC	DATA SCIENTIST	<u>51</u> 860 CHICAGO, IL	<u>2</u> 019 80912	80912
2	DSFEDERAL INC	DATA SCIENTIST	60420 ROCKVILLE, MD	<u>2</u> 019 59280	59280
3	ASCENDUM SOLUTIONS LLC	DATA SCIENTIST	72100 CINCINNATI, OH	<u>2</u> 019 94220	94220
4	TEKSYSTEMS INC	DATA SCIENTIST	80000 DEARBORN, MI	<u>2</u> 019 82820	82820
5	JACKPOCKET INC	DATA SCIENTIST	84100 NEW YORK, NY	<u>2</u> 019 102200	102200
6	TEKSYSTEMS INC	DATA SCIENTIST	90000 DEARBORN. MI	2019 82820	82820

Decision tree-



Random Forest-

```
Random Forest
24 samples
 4 predictor
No pre-processing
Resampling: Bootstrapped (25 reps)
Summary of sample sizes: 24, 24, 24, 24, 24, 24, ...
Resampling results across tuning parameters:
       RMSE
                 Rsquared
  mtry
                           MAE
   2
        35747.45 0.3557401
                           29834.16
  23
        38388.78 0.2300153
                           31804.46
       40042.45 0.1831441 33045.97
  44
RMSE was used to select the optimal model using the smallest value.
The final value used for the model was mtry = 2.
Call:
 randomForest(x = x, y = y, mtry = param$mtry, trainControl = ..1)
                 Type of random forest: regression
                        Number of trees: 500
No. of variables tried at each split: 2
           Mean of squared residuals: 1096937403
                       % Var explained: 4.22
```

Results

The analysis of the data set produced many different answers to the posed question of what salary in 2023 would be best to negotiate for with the skills and knowledge of an education for Data Science. The initial results showed an increase year over year in the mean salary offered from the different job listings. The mean salary was \$118,000.00 for all the years combined and the mean salary for 2022 was \$129,000 which is close to the mean for all the years. T can also be seen that there was around a \$5200.00 increase each year to the mean salary for the first three years and the las year showed an increase of around \$7000 to the 2022 mean salary.

The first model is a histogram of salary and frequence, this showed that the most frequent salaries were between \$100-200K with under \$100,000 being the second. Over \$200,000 made a significant drop with the least amount of reportted salaries. The density charts, and additional bar

charts showing the frequencey and desity reinfoced this with showinng the same information on where the most frequent salaries were displayed.

A word cloud was created to display the most frequenct key words used in the decription of the job listing. These key words were limited to the top 50 words for easier review and display. These showed words like Data Scientist which would be expected as the top as this is listed in every description. After those, word like machine, analytics, insight, analyst, and engeinner were displayed.

The models then move into the data mining portion of the analysis. The first is the probability table, in this all the salaries are listed and their probabilities. The most frequent probability was the 9.384384... and 1.876877... this was run on the training data. Once that was viewed the Naïve Bayes model was created to use for the prediction model. Looking at the head of the tst dataset after prediction was applied the predictions show a range of different salaries for the top 6 chosen.

A decision tree was created uing the salary and start date to determine which lead to the best salary choice. There are two branches on this tree, 2020 and under, and greater than 2020. Different tuning was done to try and create the best model and this set up created the most understandable model. The results show that the greatest salary is over the year 2021.

RandomForest was the used for the final models in the data. The data was sampled and a training and test set created. The probability table used the training data that contained the sample set. All prbabilities with the exception of one were 0.04166667. This was then applied to the training data and a model created. There were 24 samples and 4 predictors. Mtry of 2 was used as the final value as this was the lowest and determined to be the optimal model.

Conclusions

In a final review of the data, year after year the salaries for a Data Scientist increase. There was a steady increase of an average of \$5000.00 for the first three years than an increase of \$7000.00 from 2021 to 2022's salary averages. It can be seen in the average salaries per city that there are great differences in the salaries.

While there are differences in the average salaries per city, the average salary overall for all years was \$188,000 and the most frequent salary offered overall was \$120,00.00. After performing prediction analysis on the data, it can be seen in the top 6 results around a \$10,000.00 difference in what was offered and what was predicted.

When applying for a Data Science role in 2023 based on all analysis and results seen, a negotiated or expected amount for a salary shows to be between \$124,671 - \$134,671.