

Most Valued Data Science Skills

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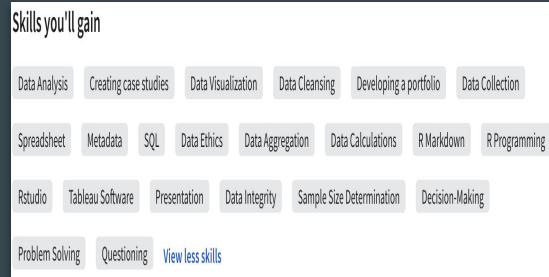
Resources

Reddit



- R Package "RedditExtractoR"
- Scrape comments from the past year
 - r/datascience (1.2m members)
 - r/DataEngineering (169k members)
- Yielded a combined data frame of over 120,000 comments for analysis.
- Project's replicability is somewhat limited due to the dynamic nature of the subreddits

Coursera



- Python lib BeautifulSoup4 and request
- Extract advertised skills that are to be acquired from most popular data science courses

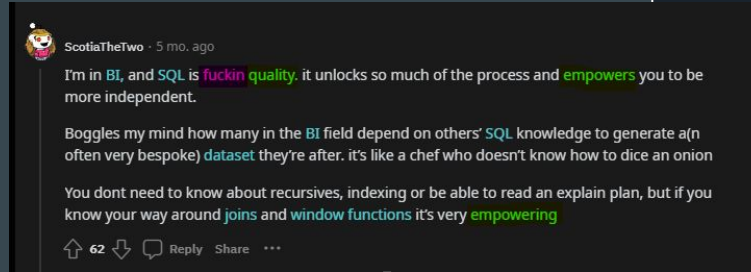
Linkedin



- Python lib BeautifulSoup4 and request
- Extract "Skills Required" section from over 50 of the latest job postings on LinkedIn

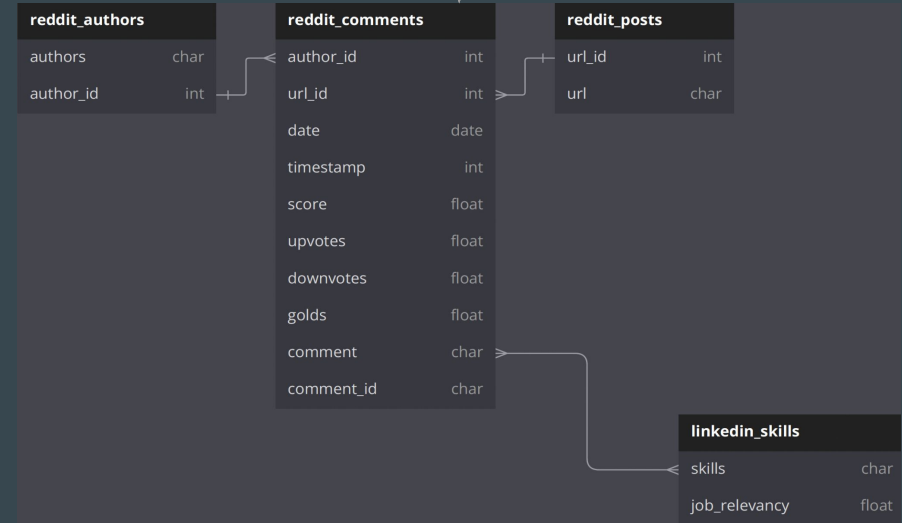
Data Loading

- Two files were created to scrape Reddit comments and LinkedIn job postings



Skills	Relevancy
Prepare Data for Exploration	3.583780
Information Technology (IT) Architecture	3.417614
Process Data from Dirty to Clean	3.064806
Process Data from Dirty to Clean	3.064806
Data Science	2.439058
Creating case studies	2.275386
Microsoft Excel	2.244409
Data Visualization	2.142001
Developing a portfolio	2.118805
Database (DBMS)	2.025451

- Data was tidied and then normalized into a SQLite database
 - reddit_comments
 - reddit_authors
 - reddit_posts
 - linkedin_skills



Sentiment Analysis Prep

Step 1 - Sentiments

AFINN by Finn Årup Nielsen, **bing** by Bing Liu and collaborators, **nrc** by Saif Mohammad and Peter Turney.

These dictionaries categorize **sentiments** (positive/negative) and **emotions** (joy/anger/disgust/etc) on a scale from -5 to 5, ranking positivity.

word	value	sentiment	word	value	sentiment
1 abandon	-2	fear	1 hurrah	5	joy
2 abandon	-2	negative	2 hurrah	5	positive
3 abandon	-2	sadness	3 outstanding	5	joy
4 abandoned	-2	anger	4 outstanding	5	negative
5 abandoned	-2	fear	5 outstanding	5	positive
6 abandoned	-2	negative	6 superb	5	positive
7 abandoned	-2	sadness	7 brilliant	4	anticipation
8 abduction	-2	fear	8 brilliant	4	joy
9 abduction	-2	negative	9 brilliant	4	positive
10 abduction	-2	sadness	10 brilliant	4	trust

Step 2 - Two-word Skills

Two-word skills were merged into single words (e.g., "**Data Science**" became "**datascience**")

We applied the same transformation to all the Reddit comments collected by Lucas using gsub.

Skills	Relevancy	word
25 Prepare Data for Exploration	3.5837803	dataexploration
65 Information Technology (IT) Architecture	3.4176142	informationtechnology
67 Process Data from Dirty to Clean	3.0648062	processdata
68 Process Data from Dirty to Clean	3.0648062	clean
49 Data Science	2.4390583	datascience
15 Creating case studies	2.2753861	casestudies
64 Microsoft Excel	2.2444093	excel
13 Data Visualization	2.1420012	visualization
6 Developing a portfolio	2.1188049	portfolio
59 Database (DBMS)	2.0254509	database

Step 3 - Separate Words

Comments were **unnested** into individual words (grouped by comment ID).

Using the sentiment dictionaries, each word was assigned a **score** and a **sentiment**.

github	0	NA
copilot	0	NA
is	0	NA
amazing	4	positive

Sentiment Analysis Results

Step 4 - Filtering for Skill words

We filtered for words that matched our Skills.

Using those comment IDs, we then filtered the full comment scrape to just include comment IDs that were in that list.

github	0	NA	Github
copilot	0	NA	NA
is	0	NA	NA
amazing	4	positive	NA

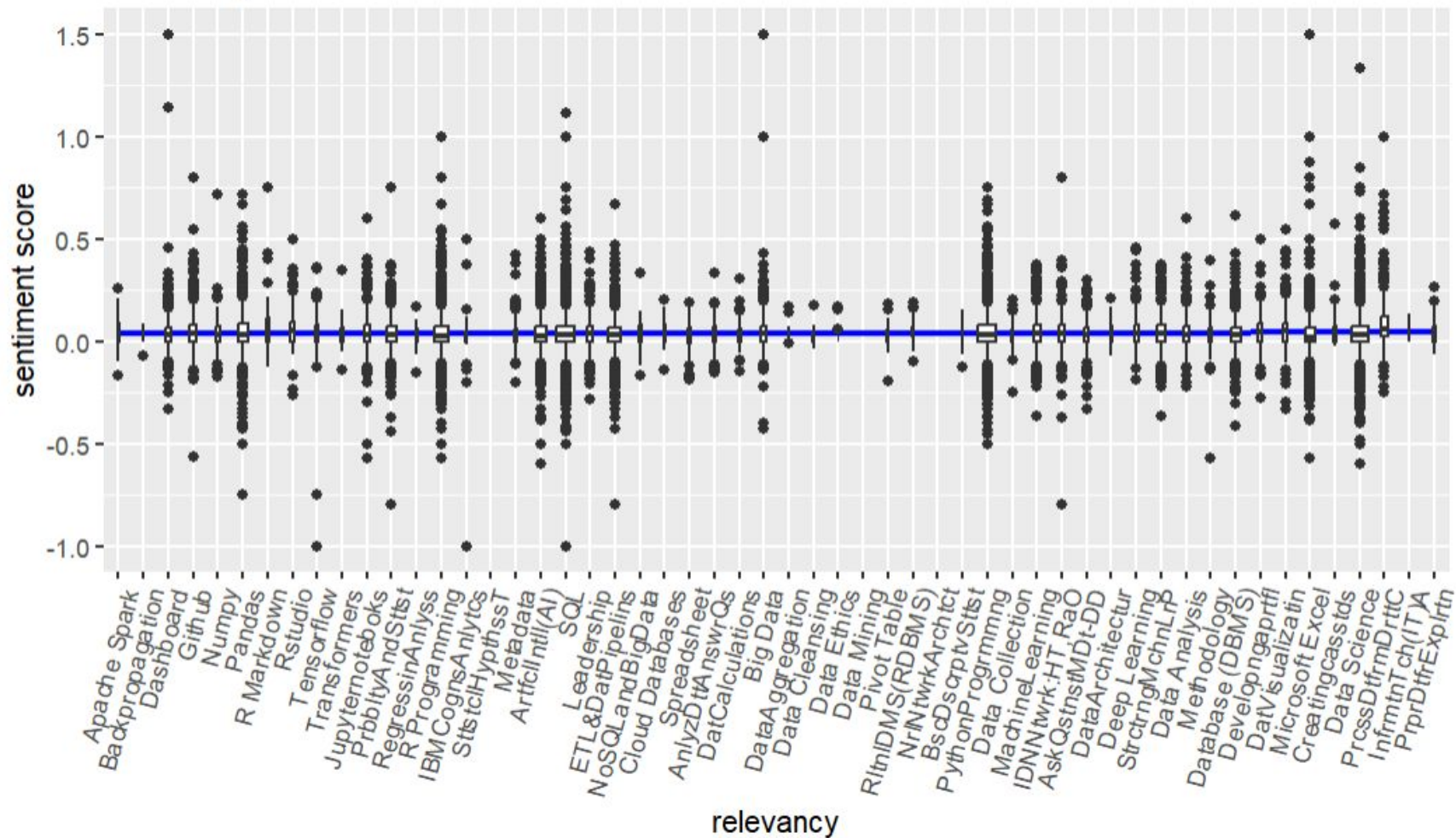
Step 5 - Average Skill Sentiment Score

We grouped words first by comment ID for **Average Comment Sentiment Score**, and then by Skills for **Average Skills Sentiment Score**.

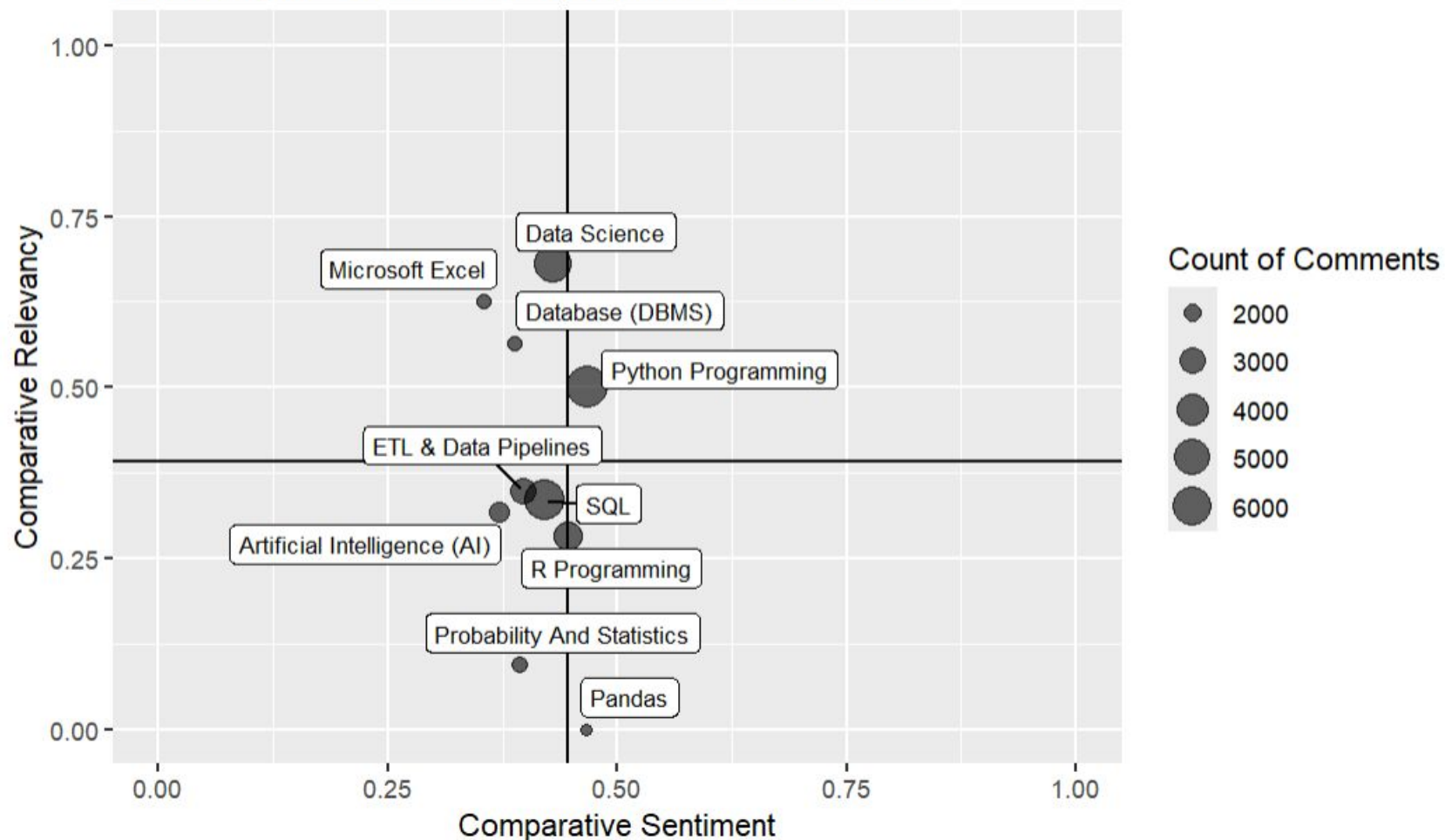
'Sum of sentiment scores in a comment' / 'Total words in a comment'
=Average Comment Sentiment Score

Skills	Relevancy	mean_score	rescaled_score	RescaledRelevancy
Process Data from Dirty to Clean	3.0648062	0.087231315	1.000000000	0.85518809
R Markdown	0.0000000	0.078117965	0.892034297	0.00000000
Data Aggregation	1.6286817	0.064467980	0.730323158	0.45445913
Rstudio	0.0000000	0.064385592	0.729347105	0.00000000
Creating case studies	2.2753861	0.064112769	0.726114974	0.63491227
Information Technology (IT) Architecture	3.4176142	0.057065673	0.642628150	0.95363386
Data Visualization	2.1420012	0.054597699	0.613390102	0.59769321
Github	0.0000000	0.052550012	0.589131184	0.00000000
Apache Spark	0.0000000	0.050742113	0.567713040	0.00000000
Big Data	1.6286817	0.050619318	0.566258287	0.45445913
Neural Network Architecture	1.7275182	0.050000000	0.558921240	0.48203798
Deep Learning	1.9331006	0.049453391	0.552445570	0.53940265
Developing a portfolio	2.1188049	0.048973881	0.546764820	0.59122066

relevancy vs. sentiment Values



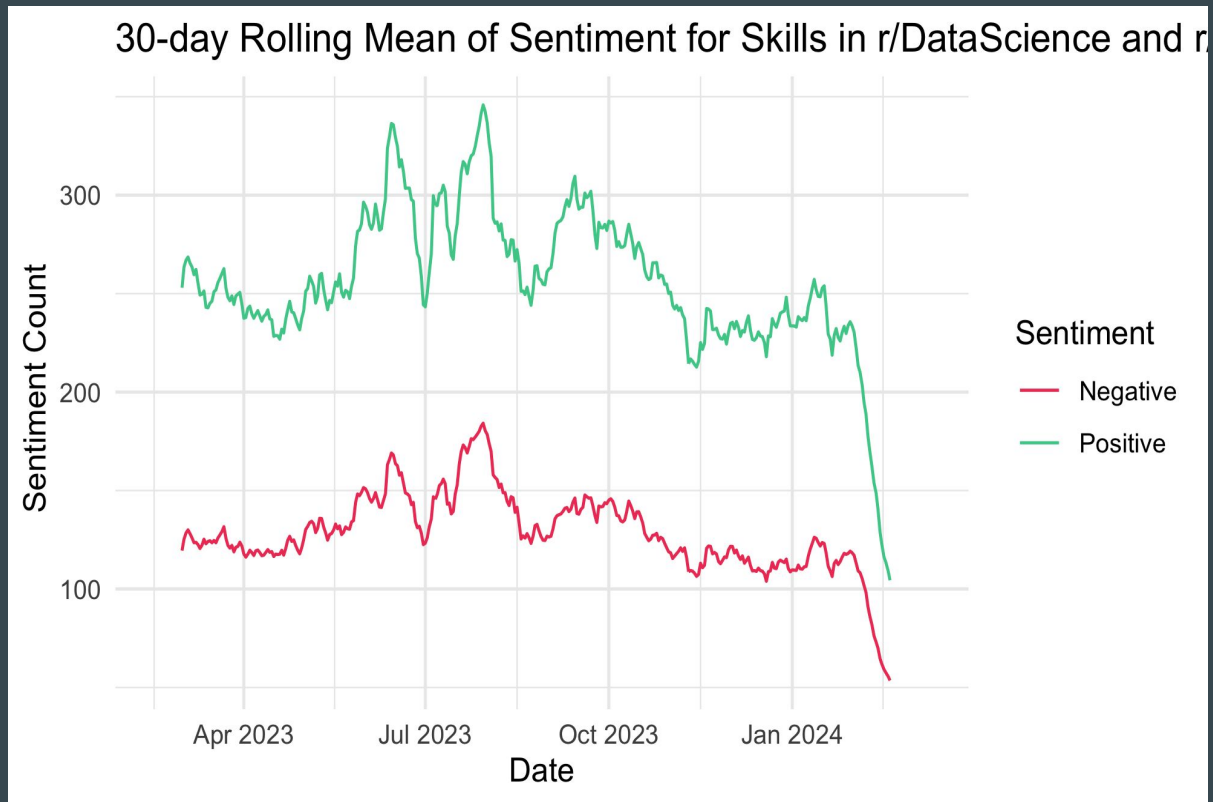
Relevancy vs Sentiment in Most Commented Skills



Results

Findings

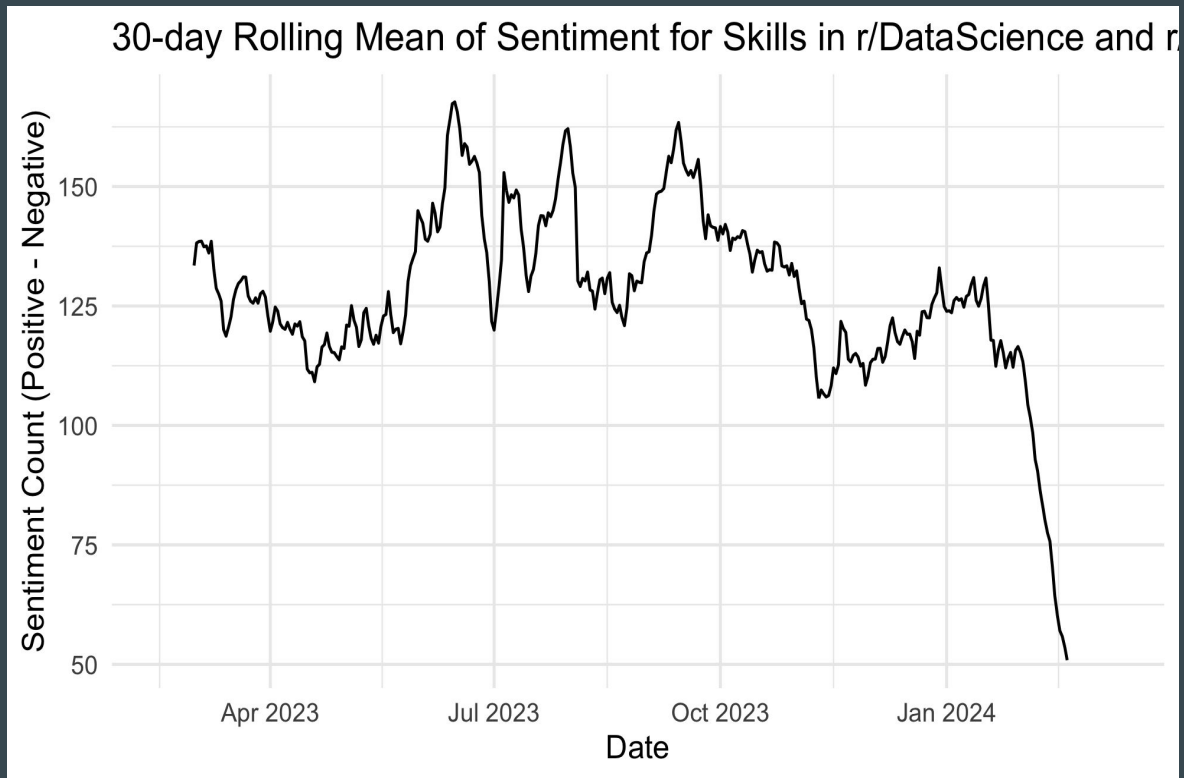
- On average there are more positive sentiments than negative sentiments in past year
- Drop off due to decreasing total comment amounts for recent comments (> 1 month)



Results

Findings

- Net sentiments usually ranges from +110 to +170



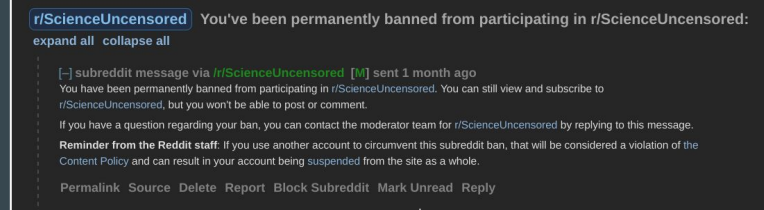
Conclusion

Data Science
Pandas SQL
Probability And Statistics
Python Programming
Artificial Intelligence (AI)
ETL & Data Pipelines
R Programming
Microsoft Excel

surprise joy
trust anger
fear
sadness
anticipation

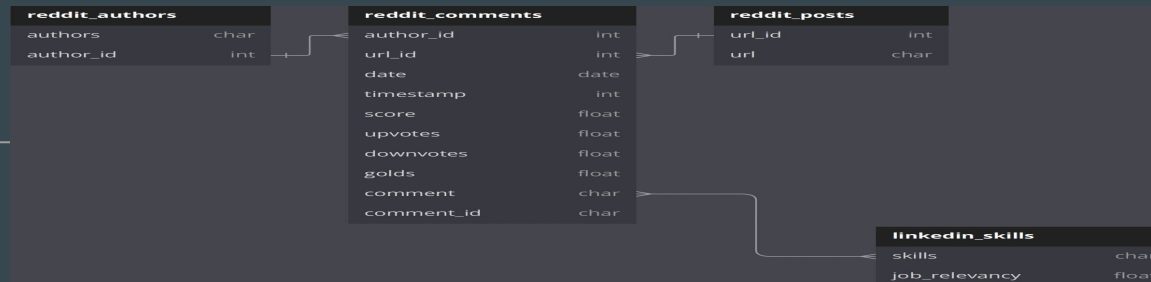
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- Skills script executed and results saved in tab delimited text file
- Reddit_scrape.r file executed and results saved in tab delimited text file
 - Dates and numeric columns converted to proper data types
 - Reddit dataset encoded to create reference tables for:
A . reddit_comments B. reddit_authors C. reddit_posts
 - Four datasets written to tables in same sqlite database:

A . reddit_comments: comments information and content • **B . reddit_authors:** author_id and author's name •
C. reddit_posts: url_id and url of the post • **D . linkedin_skills:** skills outlined in job posting and relevancy score



Analysis Methods Section 3

Step 1

- Computed 30-day rolling average time-series for sentiments
 - Show absolute numbers and differences

Data Sources & Process

- Reddit comments from the r/datascience and r/dataengineering subreddits
- What linked in jobs did we scrape?
- How did we orchestrate our scripts together and create the database file we used?
- I think Richie wrote this in the Google doc within the section where we went over how we got our data.