

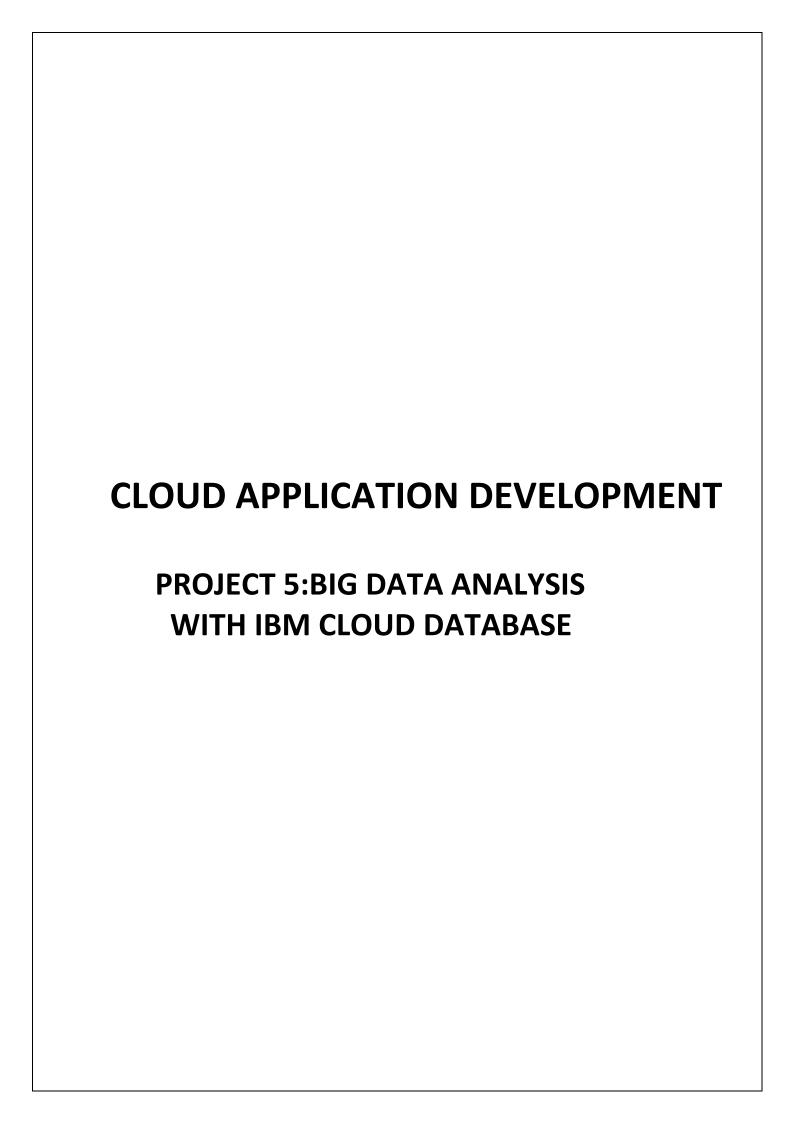
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Navigating the Path to Big Data Insights with IBM Cloud Databases

1.1. Introduction

In today's data-driven world, the ability to harness the power of big data is essential for informed decision-making and innovation. The "Big Data Analysis" project is a comprehensive exploration of data analytics using IBM Cloud

Databases. It seeks to extract valuable insights from extensive datasets, including climate trends and social patterns.

1.2. Problem Statement

The primary challenge is to delve into the world of big data analysis using IBM Cloud Databases. The goal is to uncover hidden insights within these datasets, which may include data related to climate trends or social media patterns. The project also includes designing the analysis process, setting up IBM Cloud Databases, conducting data analysis, and creating visualizations to derive essential business intelligence.

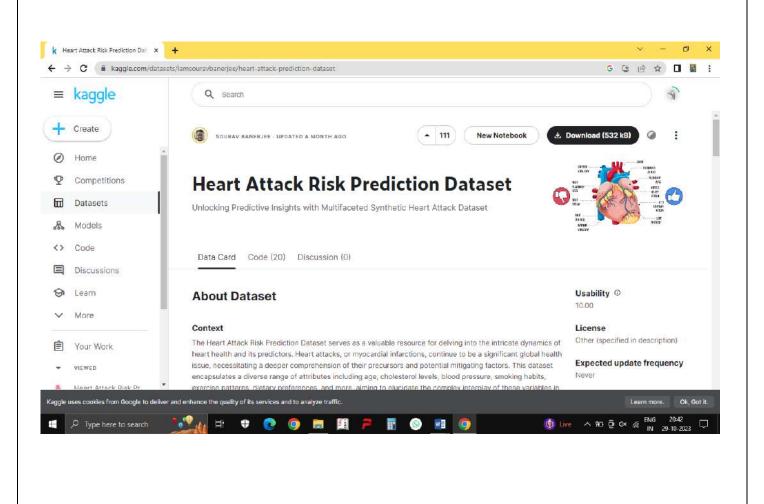
1.3. Objectives

- Identify and select relevant datasets for analysis.
- Configure IBM Cloud Databases for efficient data storage and management.
- Develop queries and scripts for data exploration.
- Apply appropriate analysis techniques, such as statistical analysis and machine learning, to extract insights.
- · Create effective visualizations to present the analysis results.
- Interpret the findings to derive actionable business recommendations.

2. Understanding the Problem Statement

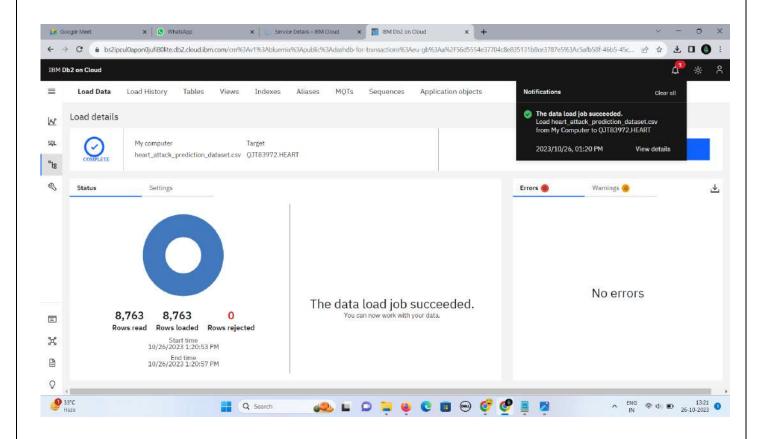
2.1. Data Selection

To address the problem statement effectively, the first step involves selecting appropriate datasets. These datasets should align with the project's objectives and can include climate data, social media trends, or other relevant data sources. The key challenge here is to ensure that the selected datasets are comprehensive and contain valuable information



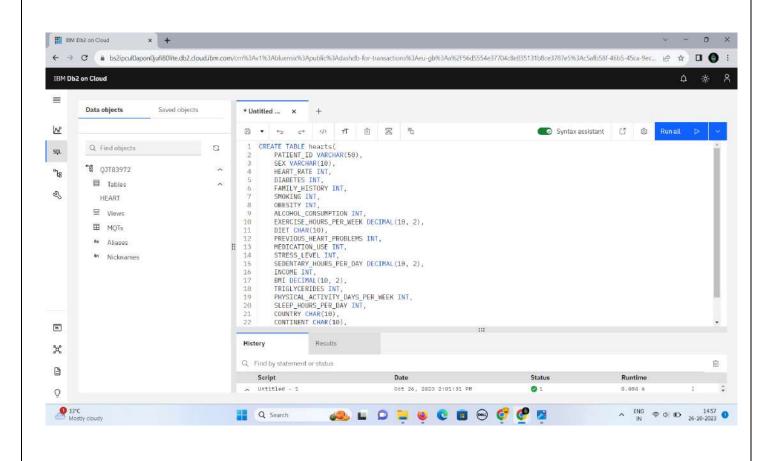
2.2. Database Setup

Efficient data management is pivotal in big data analysis. We will set up IBM Cloud Databases to store and manage the selected datasets. This step demands a keen understanding of the database infrastructure and proper configuration to handle large volumes of data effectively



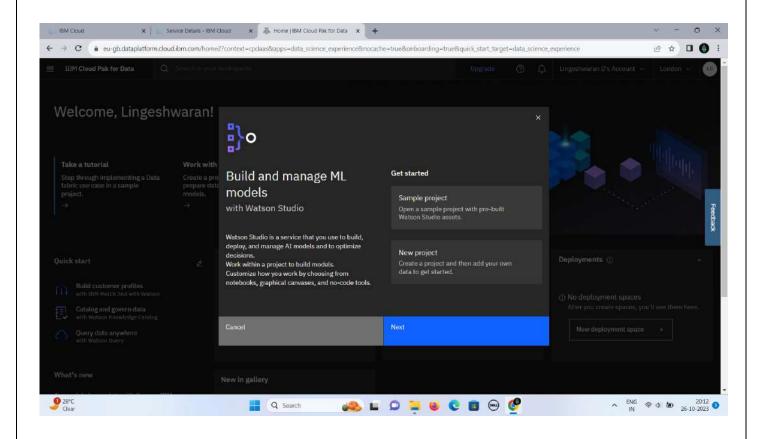
2.3. Data Exploration

The heart of this project lies in data exploration. This phase involves developing queries and scripts to delve into the datasets, extract relevant information, and identify patterns. It requires a thorough understanding of the datasets and an ability to navigate through extensive data efficiently.



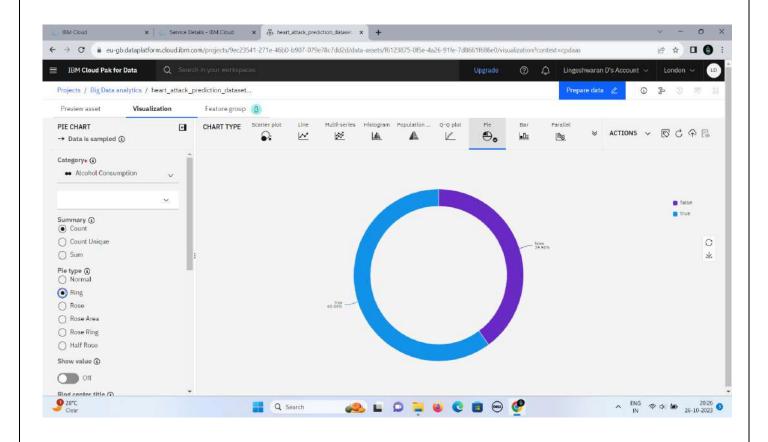
2.4. Analysis Techniques

The chosen datasets may vary in complexity, and different analysis techniques may be needed to extract insights. We will apply appropriate techniques, such as statistical analysis or machine learning, depending on the nature of the data. It's crucial to have a strong grasp of these techniques to ensure meaningful results.



2.5. Visualization

Effective communication of results is essential. To this end, we will design visualizations that not only present the analysis findings but also make them understandable and impactful. This phase demands creative and data visualization skills.



2.6. Business Insights

Business insights derived from big data analytics can provide organizations with valuable information and opportunities for informed decision-making. Here are some key insights and benefits that businesses can gain from big data analytics:

1. Customer Insights:

* Customer behavior analysis:

Big data analytics can help businesses understand customer preferences, buying patterns, and trends, allowing for more personalized marketing and product offerings.

*Customer segmentation:

Businesses can identify different customer segments based on demographics, behavior, and preferences, allowing for targeted marketing and product development.

* Churn prediction:

Analytics can help predict when customers are likely to leave, enabling businesses to take proactive measures to retain them.

2. Operational Efficiency:

*Process optimization:

Big data analytics can identify inefficiencies in business processes and suggest improvements to reduce costs and increase productivity.

*Supply chain optimization:

Analyzing data can help optimize the supply chain, reduce inventory costs, and improve order fulfillment.

3. Product Development:

*Market research:

Big data can provide insights into market trends and consumer needs, helping businesses create products and services that are in demand.

*Product quality and performance:

Analyzing customer feedback and product data can help improve product quality and performance.

4. Fraud Detection:

Anomaly detection:

Big data analytics can detect unusual patterns and anomalies in transactions, helping identify and prevent fraudulent activities.

5. Predictive Maintenance:

Equipment health monitoring:

By analyzing data from sensors and IoT devices, businesses can predict when equipment is likely to fail, reducing downtime and maintenance costs.

6. Financial Analysis:

Risk assessment:

Big data analytics can assess financial risks by analyzing data from various sources, helping organizations make informed investment decisions.

Credit scoring:

Businesses can use big data to improve credit scoring models, leading to better lending decisions.

7. Marketing and Advertising:

*Campaign effectiveness:

Analyzing marketing campaigns can help businesses understand which strategies are most effective and optimize their marketing budgets.

*Social media sentiment analysis:

Monitoring social media can provide insights into customer sentiment and public opinion.

8. Competitive Analysis:

*Market positioning:

Businesses can use big data to understand their position in the market and identify opportunities to gain a competitive edge.

*Competitor analysis:

Analyzing data on competitors can help organizations make informed strategic decisions.

9. Human Resources:

*Employee engagement:

Analytics can be used to measure and improve employee engagement and retention.

*Recruitment and talent management:

Big data can help identify the best candidates for job openings and assess employee performance.

10. Risk Management:

*Identifying and mitigating risks: Big data analytics can help organizations identify potential risks, assess their impact, and develop strategies for risk mitigation.

3. Development Phases:

The project is divided into two primary development phases.

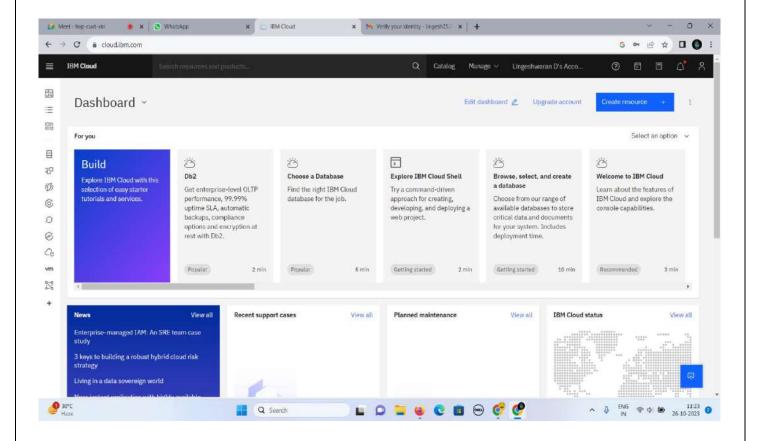
3.1. Development Part 1:

In this phase, we will initiate the big data analysis solution using IBM Cloud Databases. We will import the selected datasets, execute queries and scripts, and start to unveil initial insights.

This part contain's several steps.

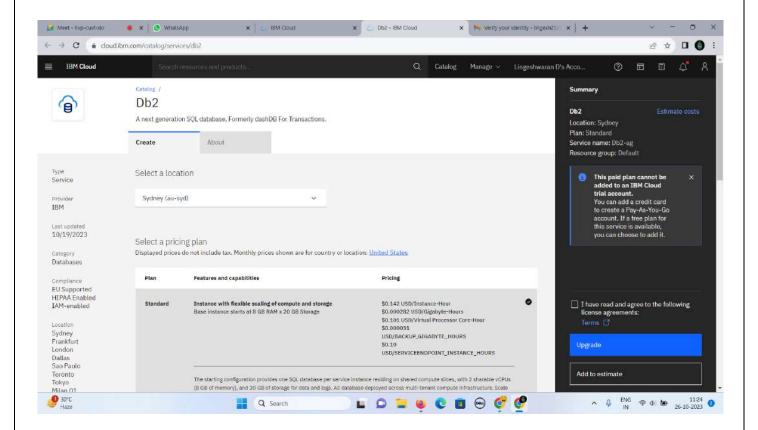
Step 1:

- *Login our IBM Cloud Account
- *Search Db2



Step 2:

- *Create our free Database 2 server
- *Select a location



Step 3:

- *Select a location london(eu-gb)
- *Create our Database 2 server

Figure 1:

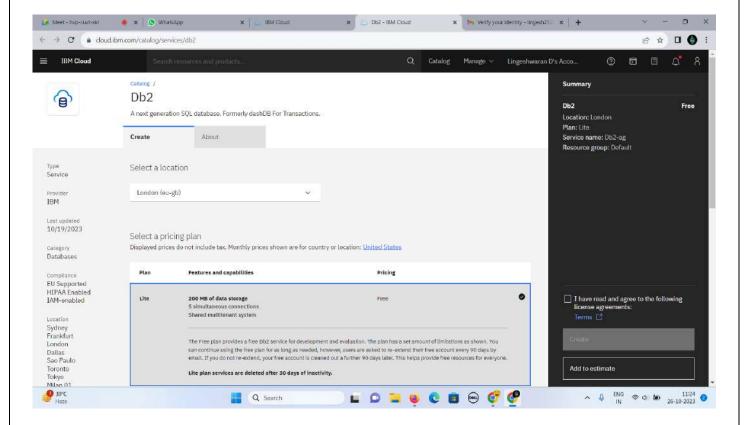
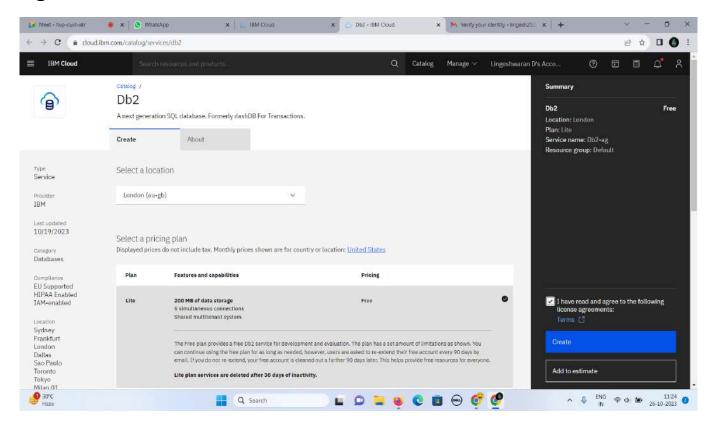


Figure 2:



Step 4:

*Goto dashboard

*select your Database server 2(Db2-ag)

Figure 1:

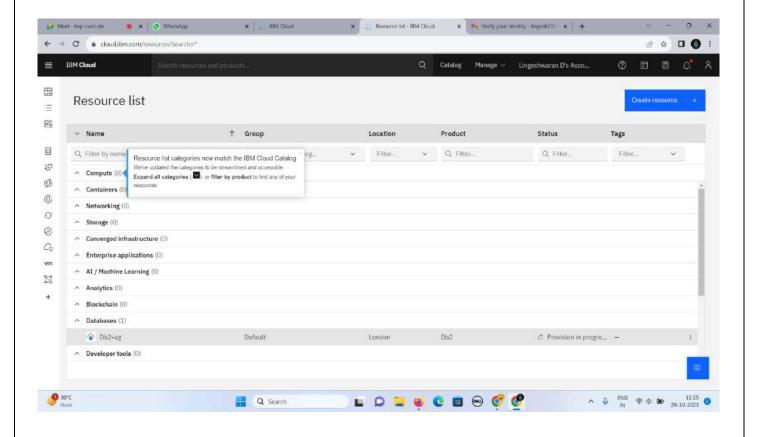
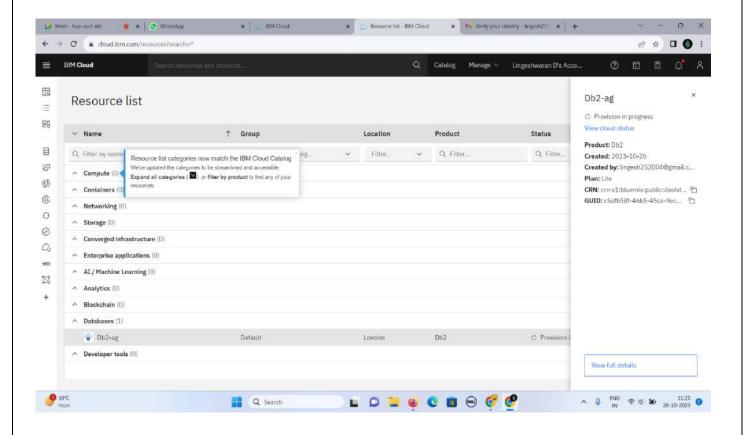
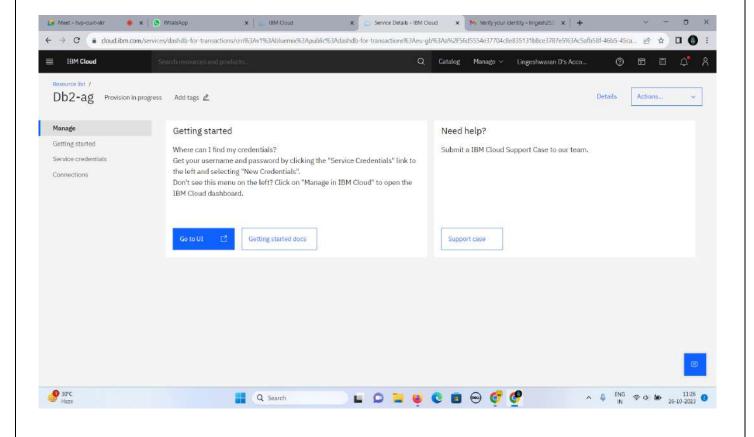


Figure 2:

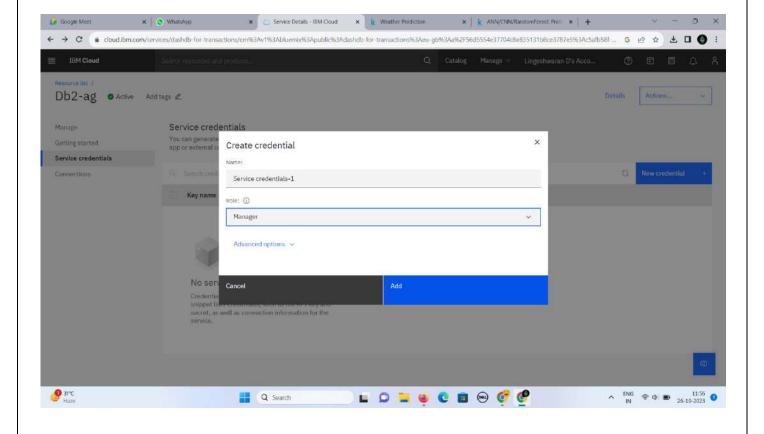


Db2-ag Interface in IBM Cloud

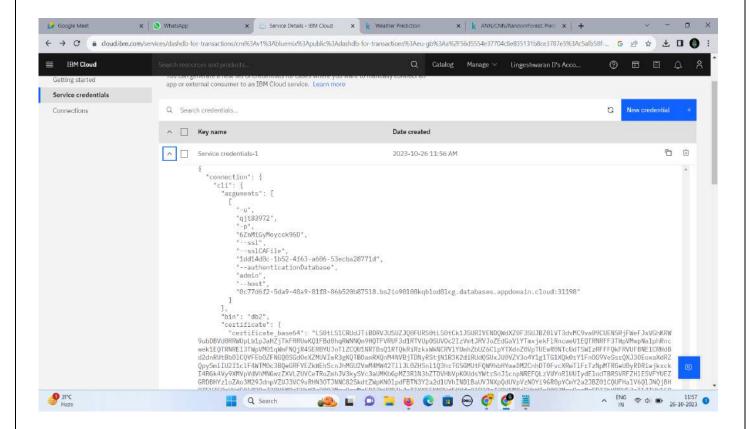


Step 5:

*Create your Server Credential-1

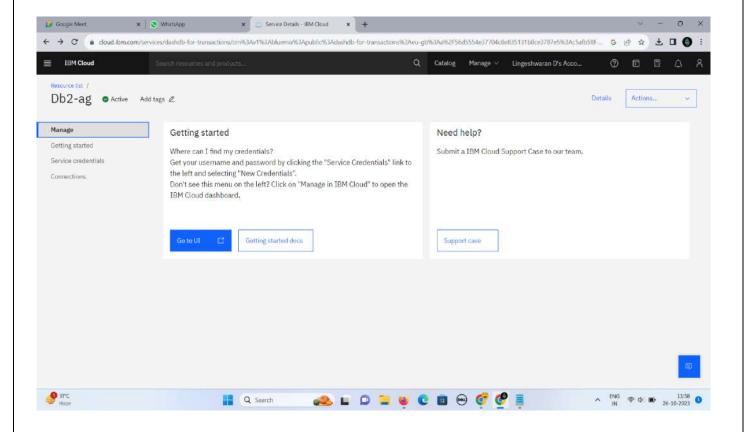


*After create server credential, then it show's our UserID, Password, Host Address

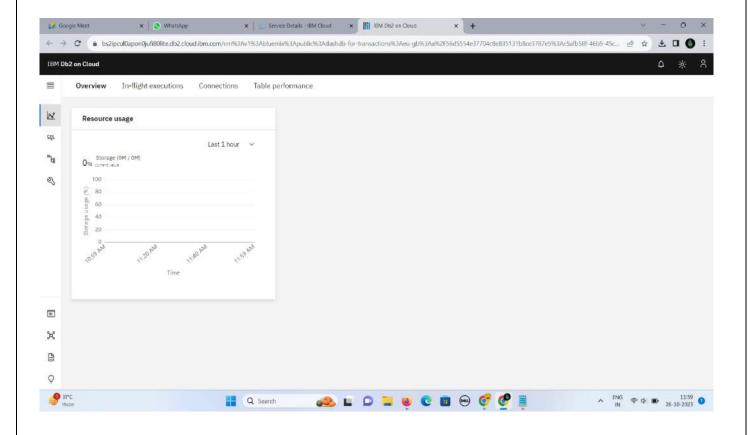


Step 6:

*click "Go To UI" for your data file upload

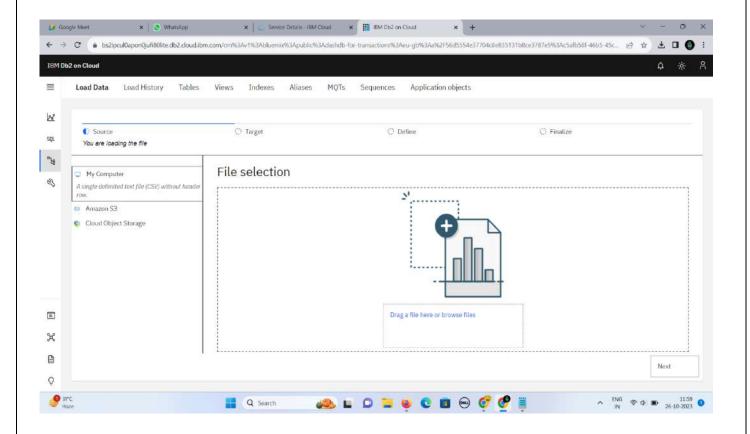


*select your Load Data

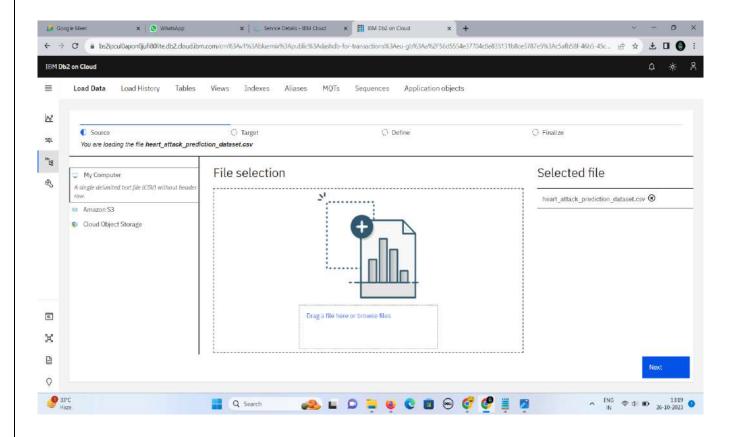


Step 7:

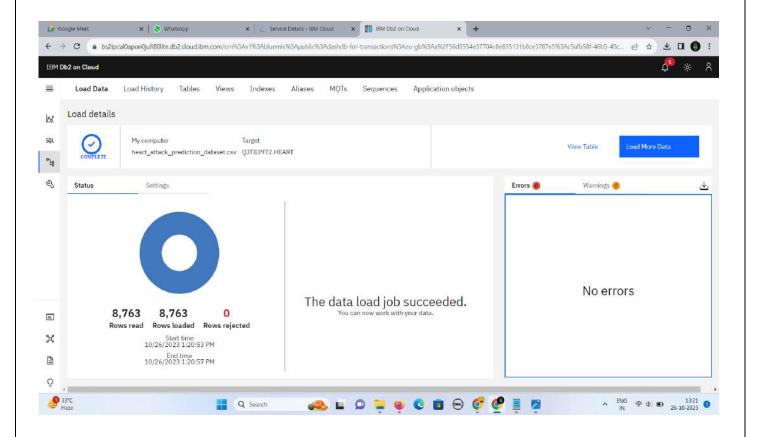
*uplaod our csv file data from "My Computer"



*upload our csv file ("heart_attack_prediction_dataset.csv")



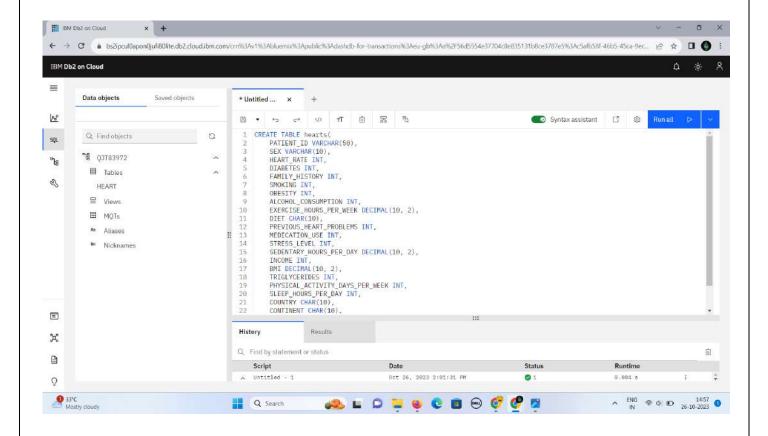
*The server read your data completely and upload your data on your server and completing your data upload process



Step 8:

*After completely uploading our dataset

*Go To SQL section and execute Data Cleaning and transformation queries



3.2. Development Part 2

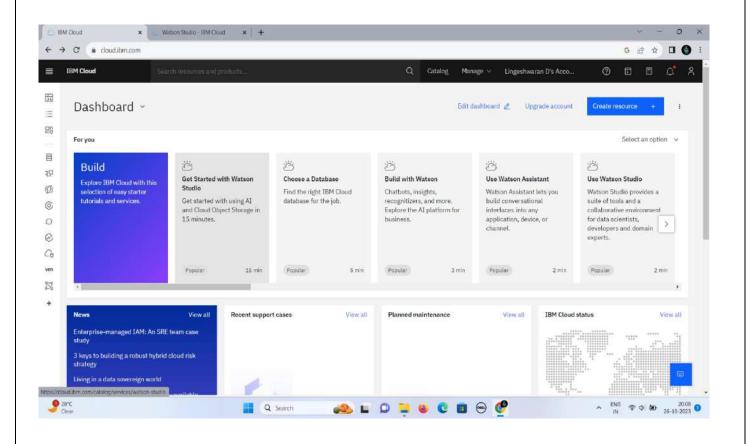
Building on the foundations laid in the first phase, we will deepen our analysis. This phase includes applying advanced analysis techniques and creating more comprehensive visualizations. It's where the true power of big data analysis comes to the forefront.

This part contains several steps.

Step 1:

*login our IBM Cloud Account and create ibm watson studio

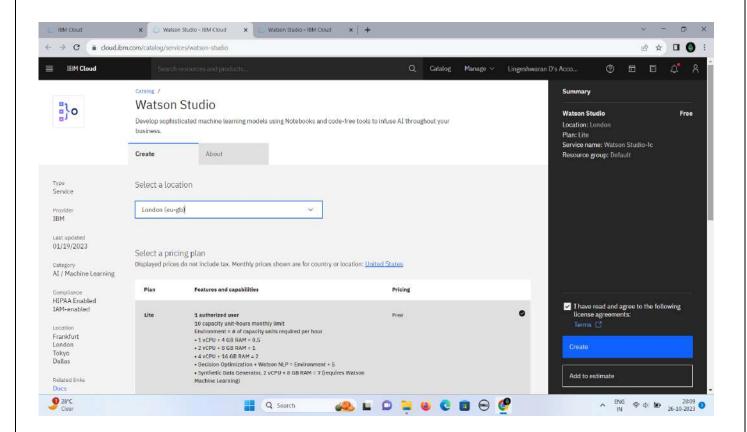
*search IBM Watson studio on search bar



Step 2:

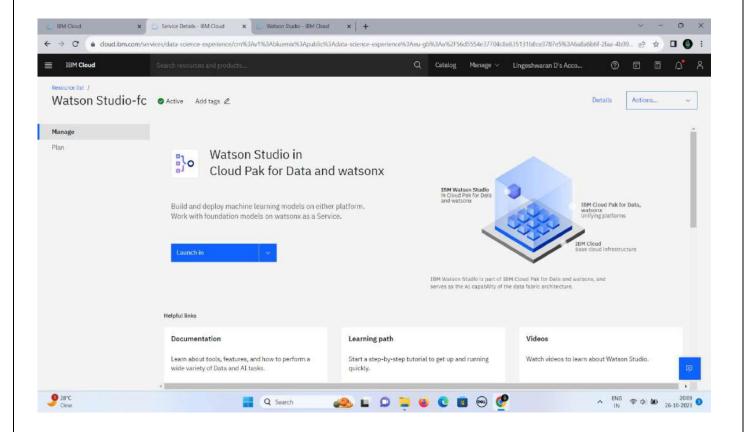
- *Create a watson studio and select a location
- *After selecting the location and click create

Figure 1:

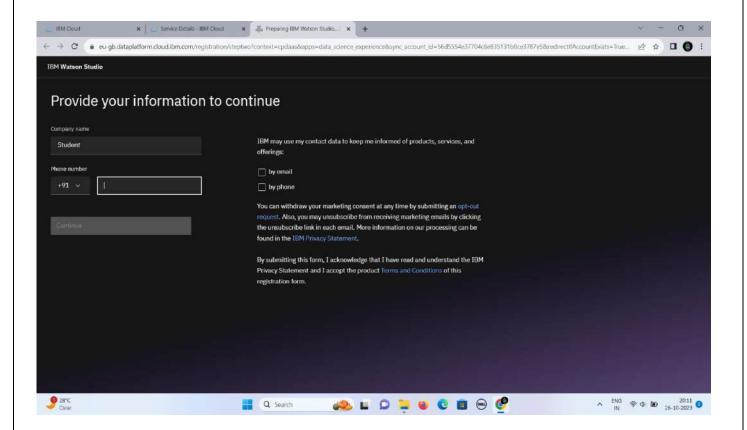


*launch the IBM Watson Studio

Figure 2:

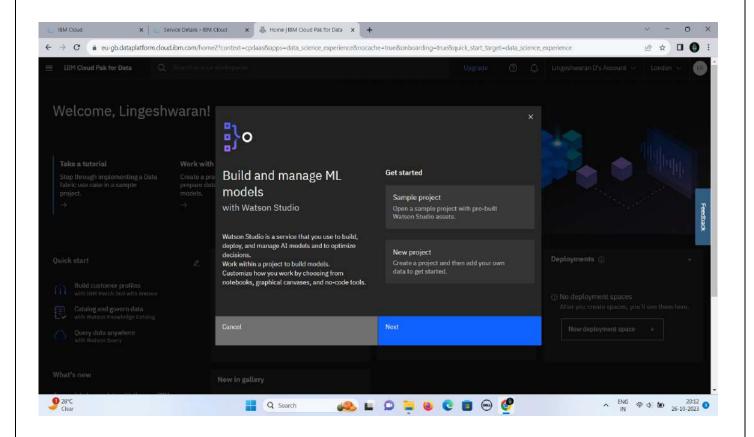


*Enter your information to continue

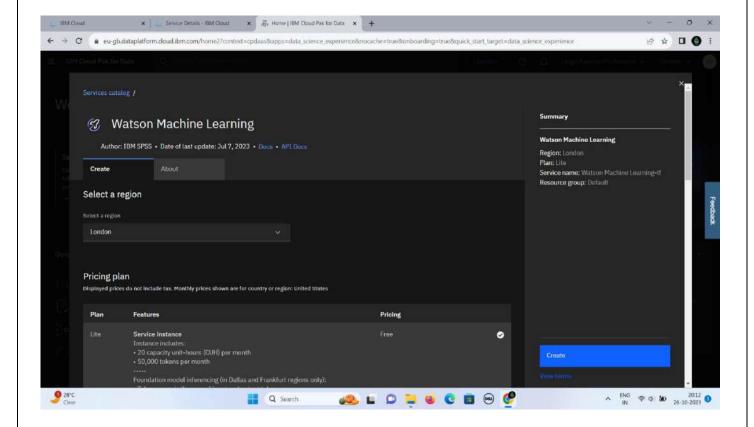


Build and manage ML model with watson studio

Watson studio is a service that you use to build, deplay and manage AI model and to optimise decision work within a project to build model customize how you work by choosing from notebook, graphical canvases and no code tools

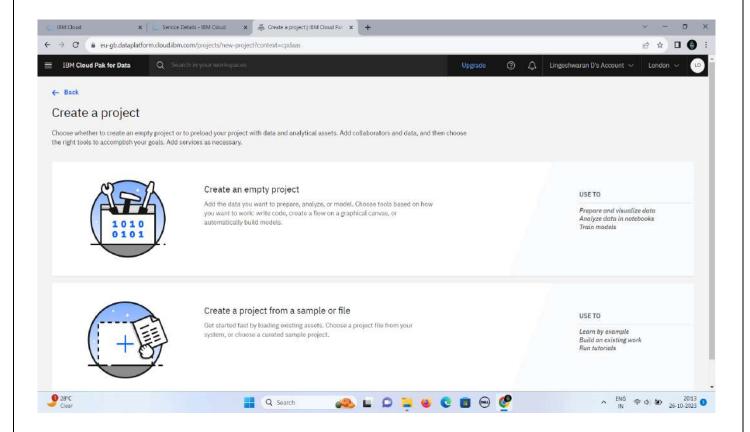


*create watson machine learning in IBM watson studio

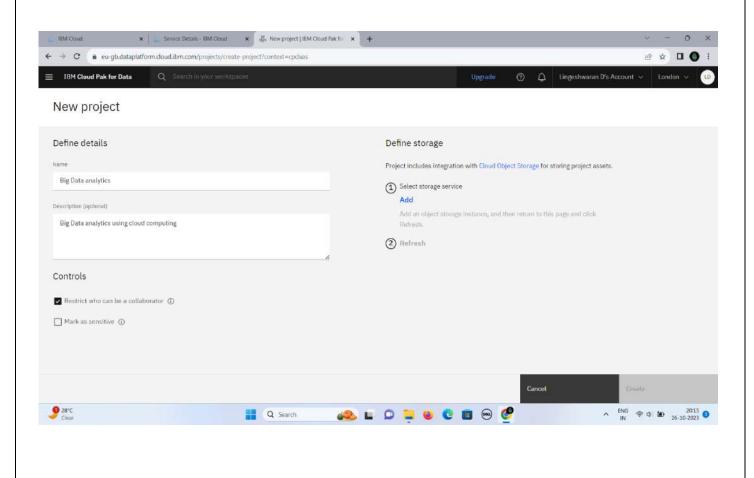


*create a project to work on IBM watson studio

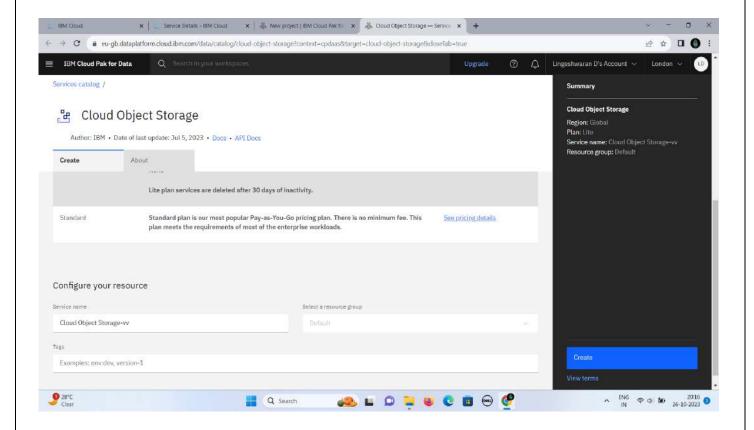
Choose whether to create an empty project or to preload with date and analytical assets.add collaborators and date, and then choose the right tools to accomplish your goals.add service as necessary



- *Give your project name and description of your project
- *Add free storage on IBM watson studio for our dataset ("Cloud Object Storage")



*"cloud object storage" purchase process for storing a dataset



* After complete a name ,description & cloud object storage process and create your project

Figure 1:

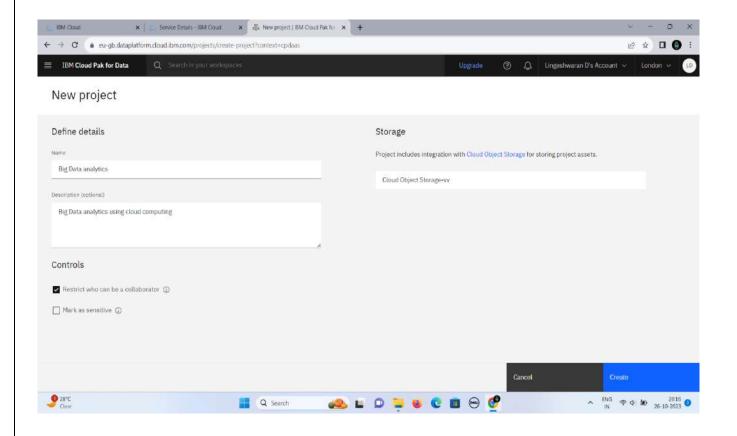
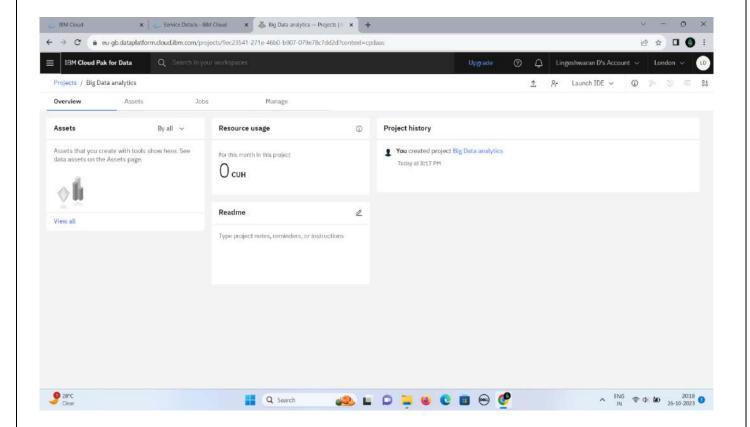


Figure 2:

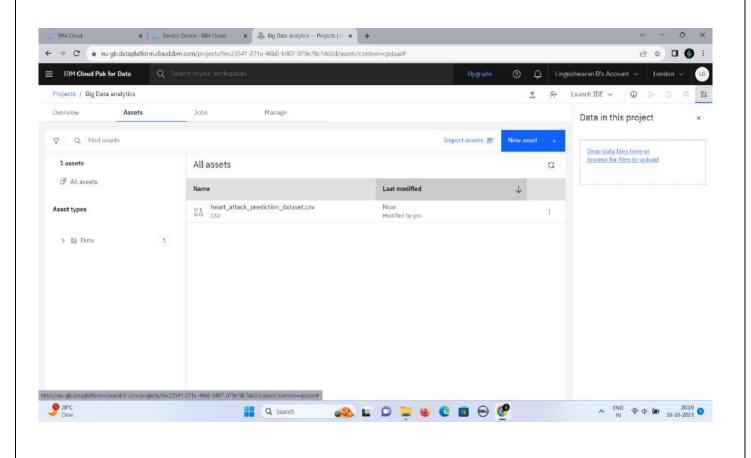


*Upload your csv dataset file on IBM watson studio ("heart_attack_prediction_dataset.csv")

*click Assets to view our csv file

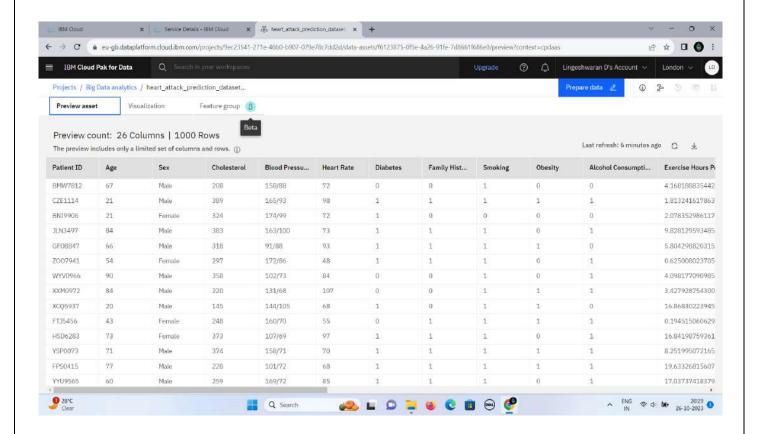
*click on your csv file for preview asset

Figure 3:



*select a visualizaiton on your csv.file

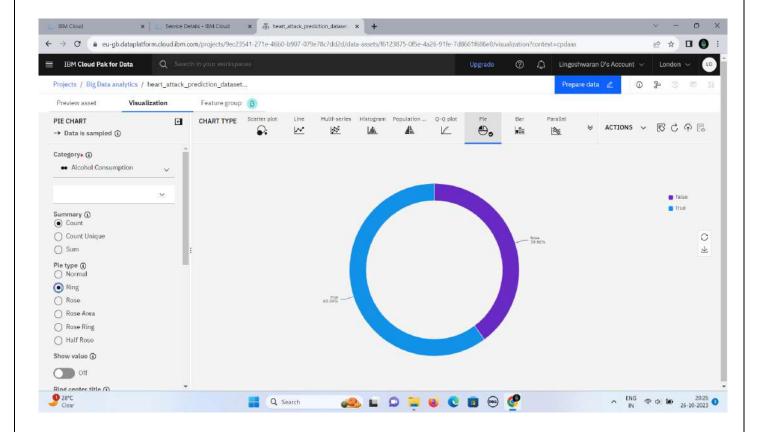
Figure 4:



Step 3:

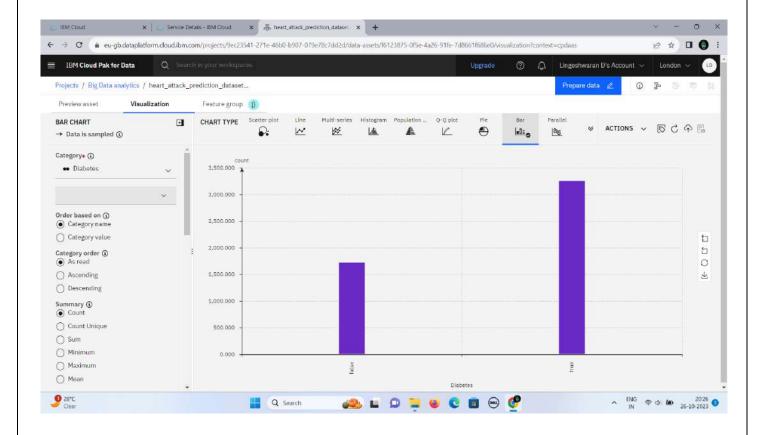
Visualization our dataset using IBM Watson studio

PIECHART:



BAR CHART:

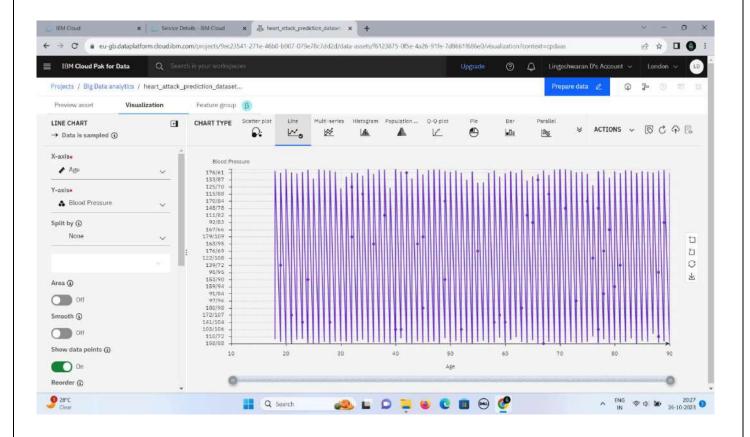
*For Diabetes



LINE CHART:

*X-AXIS(Age)

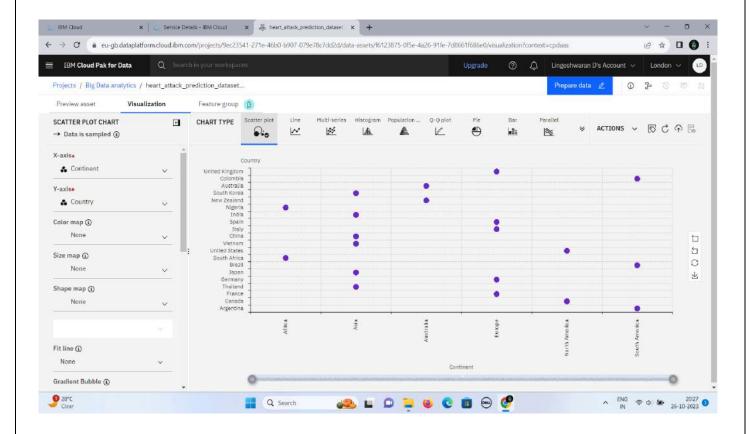
*Y-AXIS(Blood pressure)



SCATTER PLOT CHART:

*X-AXIS(CONTINENT)

*Y-AXIS(COUNTRY)

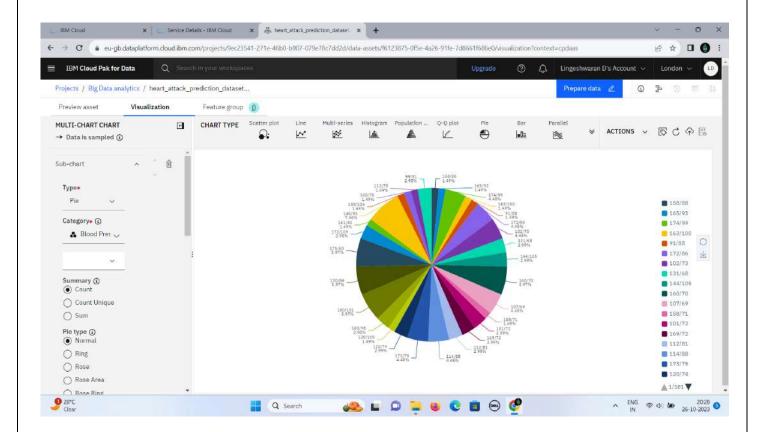


MULTI_CHART CHART:

SUB-CHART

*CATEGORY=BLOOD PRESSURE

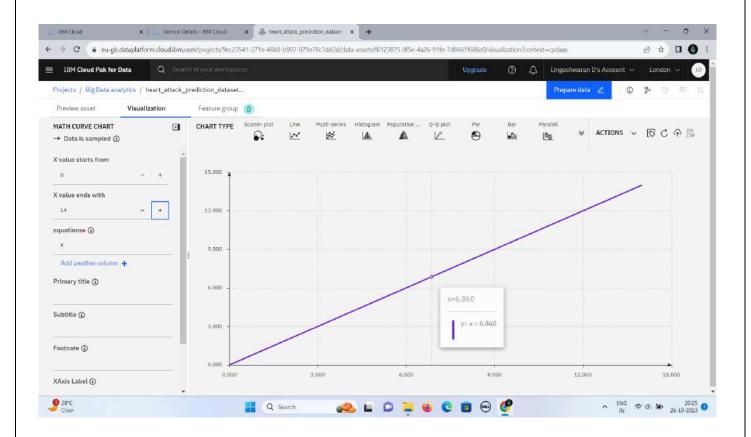
*TYPE=PIE



MATH CURVE CHART:

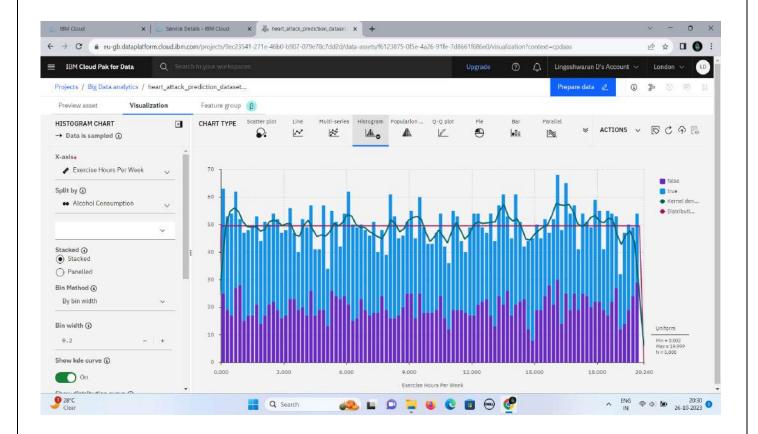
*X-AXIS(0)

*Y-AXIS(14)



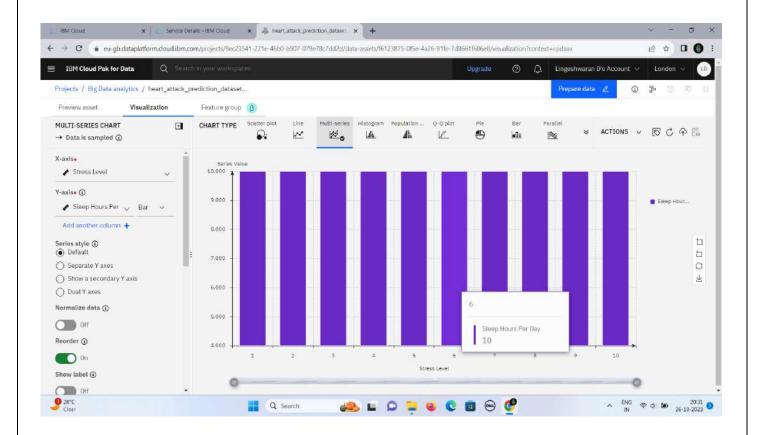
HISTOGRAM CHART:

- *X-AXIS(EXERCISE HOURS PER WEEK)
- *SPLIT BY(ALCOHOL CONSUMPTION)



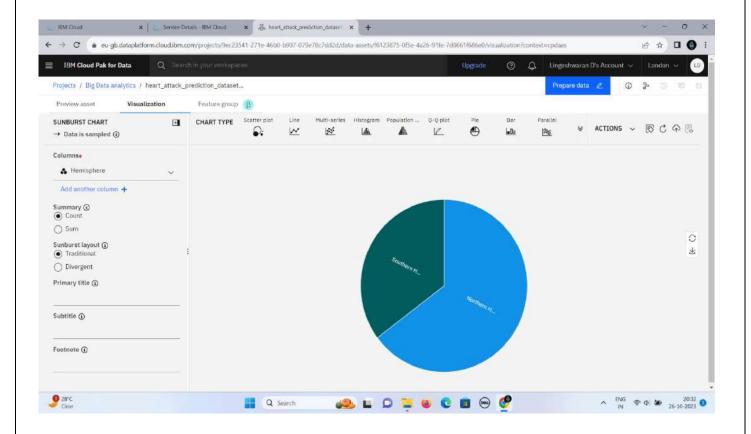
MULTI-SERIES CHART:

- *X-AXIS(STRESS LEVEL)
- *Y-AXIS(SLEEP HOURS PER WEEK)



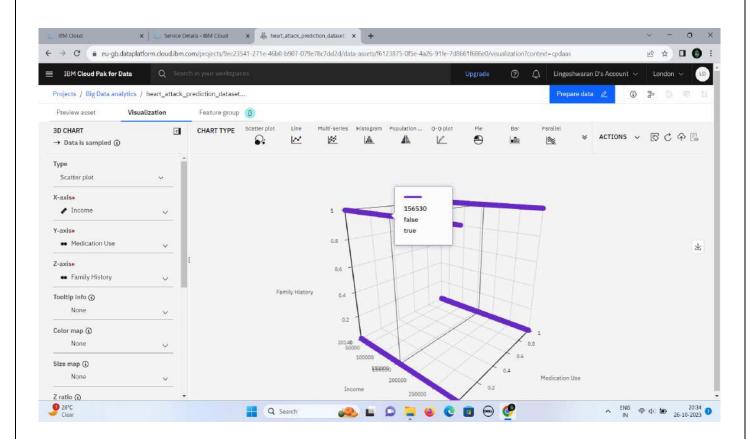
SUNBURST CHART:

COLUMNS=HEMISPHERE



3D CHART:

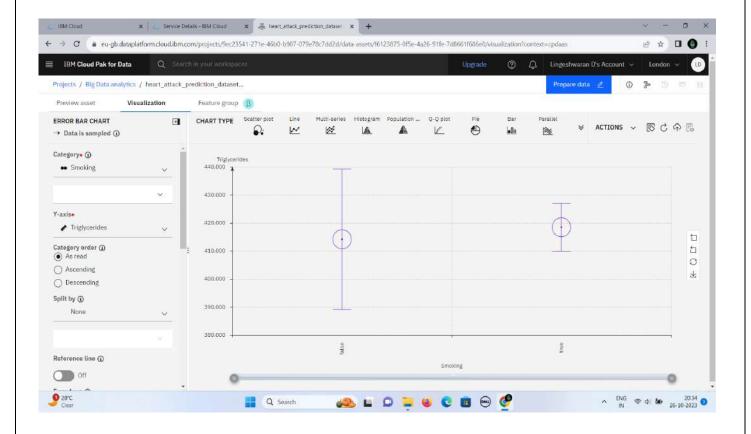
- *TYPE()
- *X-AXIS(INCOME)
- *Y-AXIS(MEDICINE USE)
- *Z-AXIS(FAMILY HISTORY)



ERROR BAR CHART:

*CATEGORY(SMOKING)

*Y-AXIS(TRIGLYCERIDES)



Big data analytics solution by applying advanced analysis techniques using python code

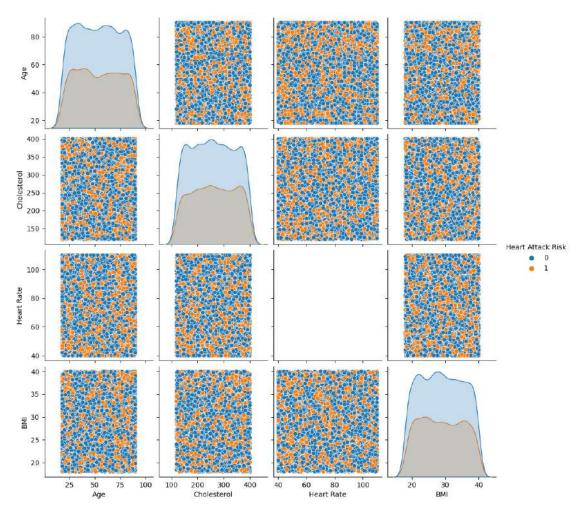
IBM cloud notebook:

```
!pip install pandas scikit-learn
Requirement already satisfied: pandas in
/opt/conda/envs/Python-3.10/lib/python3.10/site-packages (1.4.3)
Requirement already satisfied: scikit-learn in /opt/conda/envs/Python3.10 /lib/python3.10/site-packages (1.1.1)
Requirement already satisfied: python-dateutil>=2.8.1 in
/opt/conda/envs/Python-3.10/lib/python3.10/site-packages (from pandas) (2.8.2)
Requirement already satisfied: pytz>=2020.1 in /opt/conda/envs/Python3.10 /lib/python3.10/site-packages (from
pandas) (2023.3.post)
Requirement already satisfied: numpy>=1.21.0 in
/opt/conda/envs/Python-3.10/lib/python3.10/site-packages (from pandas)(1.23.1)
Requirement already satisfied: threadpoolctl>=2.0.0 in
/opt/conda/envs/Python-3.10/lib/python3.10/site-packages (from scikitlearn) (2.2.0)
Requirement already satisfied: joblib>=1.0.0 in
/opt/conda/envs/Python-3.10/lib/python3.10/site-packages (from scikitlearn) (1.1.1)
Requirement already satisfied: scipy>=1.3.2 in /opt/conda/envs/Python-
3.10 /lib/python3.10/site-packages (from scikit-learn ) (1.8.1)
Requirement already satisfied: six>=1.5 in /opt/conda/envs/Python3.10/lib/python3.10/site-packages (from python-
dateutil>=2.8.1> pandas ) (1.16.0)
#loading the dataset
import os, types import pandas as pd from
botocore.client import Config import ibm_boto3 def
__iter__(self): return 0
# @hidden_cell
# The following code accesses a file in your IBM Cloud Object Storage.
It includes your credentials.
# You might want to remove those credentials before you share the notebook. cos_client =
ibm_boto3.client(service_name='s3',
  ibm_api_key_id='bNtKn-suJyy3PdlnIbNENRb7ZoRxjw0Bk6tjKdfwXk9H',
ibm auth endpoint="https://iam.cloud.ibm.com/oidc/token", config=Config(signature version='oauth'),
endpoint_url='https://s3.private.eu-gb.cloud-objectstorage.appdomain.cloud')
bucket = 'bigdataanalyticsusingcloudcomputi-donotdelete-prbnssfschsf5bmp'
object_key = 'heart_attack_prediction_dataset.csv'
body = cos client.get object(Bucket=bucket,Key=object key)['Body'] # add missing iter method, so pandas
accepts body as file-like object
```

```
df = pd.read_csv(body)
df.head()
  Patient ID
                             Cholesterol Blood Pressure Heart Rate
                                                                        Diabetes
               Age
                       Sex
0
     BMW7812
                      Male
                                      208
                                                   158/88
                67
                                                                    72
1
                21
                      Male
                                      389
                                                                    98
                                                                                1
     CZE1114
                                                   165/93
2
     BNI9906
                21
                    Female
                                      324
                                                  174/99
                                                                    72
                                                                                1
3
                                                                    73
     JLN3497
                84
                      Male
                                      383
                                                  163/100
                                                                                1
4
     GF08847
                      Male
                                      318
                                                    91/88
                                                                    93
                                                                                1
                66
   Family History
                    Smoking
                              Obesity
                                       . . .
                                             Sedentary Hours Per Day
                                                                        Income
0
                                                                        261404
                                                             6.615001
                 0
                           1
                                     0
                                        . . .
                                                                        285768
1
                 1
                           1
                                     1
                                                             4.963459
                                        . . .
2
                 0
                           0
                                                             9.463426
                                                                        235282
                                     0
                                        . . .
3
                 1
                           1
                                     0
                                                             7.648981
                                                                        125640
                                        . . .
4
                 1
                           1
                                     1
                                                             1.514821
                                                                        160555
                                        . . .
                               Physical Activity Days Per Week
               Triglycerides
   31.251233
0
                          286
   27.194973
                          235
                                                                1
1
2
   28.176571
                          587
                                                                4
3
   36.464704
                          378
                                                                3
4
   21.809144
                          231
                                                                1
   Sleep Hours Per Day
                            Country
                                          Continent
                                                                Hemisphere
0
                          Argentina
                                     South America
                                                      Southern Hemisphere
                      6
                      7
1
                             Canada
                                     North America
                                                      Northern Hemisphere
2
                      4
                             France
                                             Europe
                                                      Northern Hemisphere
3
                      4
                             Canada
                                     North America
                                                      Northern Hemisphere
4
                      5
                           Thailand
                                               Asia
                                                      Northern Hemisphere
   Heart Attack Risk
0
1
                    0
2
                    0
3
                    0
4
                    0
[5 rows x 26 columns]
df.isnull().sum()
Patient ID
                                     0
                                     0
Age
Sex
                                     0
Cholesterol
                                     0
Blood Pressure
                                     0
Heart Rate
                                      0
```

Diabetes Family History Smoking Obesity Alcohol Consumption Exercise Hours Per Week Diet Previous Heart Problems Medication Use Stress Level Sedentary Hours Per Day Income BMI Triglycerides Physical Activity Days Per Week Sleep Hours Per Day Country Continent Hemisphere Heart Attack Risk dtype: int64 df_data_1.isnull().sum()	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
Patient ID	0
Age	0
Sex	0
Cholesterol	0
Blood Pressure	0
Heart Rate	0
Diabetes	0
Family History	0
Smoking	0
Obesity	0
Alcohol Consumption	0
Exercise Hours Per Week	0
Diet	0
Previous Heart Problems	0
Medication Use	0
Stress Level	0
Sedentary Hours Per Day	0
Income	0
BMI	0
Triglycerides	0
Physical Activity Days Per Week	0
Sleep Hours Per Day	0
Country	0
Continent	0
Hemisphere	0

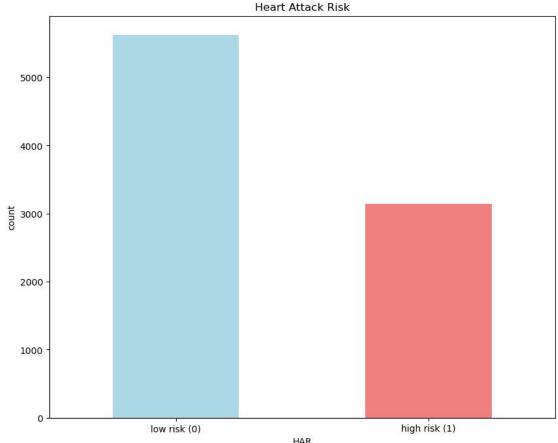
<Figure size 1000x600 with 0 Axes>



```
numeric_data =df.select_dtypes(include=["int64","float64"])
correlation_matrix = numeric_data.corr()
plt.figure(figsize=(10,8))
sns.heatmap(correlation_matrix, annot = True, cmap = 'coolwarm', fmt = ".2f")
plt.title("correlation")
plt.show()
```

correlation 1.0 Age -1.00-0.01-0.00-0.010.01 0.39-0.01-0.010.00 0.00 0.00 0.02 0.02-0.00-0.000.00 0.00-0.000.01 Cholesterol -0.011.00 0.00-0.01-0.020.02-0.010.010.02-0.010.00-0.020.02 0.00 0.02-0.010.02 0.00 0.02 - 0.8 Family History -0.01-0.02-0.01-0.0011.00 0.01-0.000.01-0.01-0.000.00 0.02 0.00-0.00-0.01-0.000.01-0.01-0.00 Smoking -0.39 0.02-0.010.00 0.01 1.00 0.00 0.01-0.000.01-0.000.02 0.00 0.01 0.00-0.01-0.01-0.00 Obesity -0.01-0.010.01 0.01-0.000.00 1.00-0.020.00 0.01-0.010.01-0.000.00-0.010.00 0.01-0.010.01 Alcohol Consumption -0.01-0.010.00 0.01 0.01 0.01-0.021.00-0.010.01 0.00-0.01-0.01-0.01-0.02 0.01 0.00 0.00 0.00 0.00 - 0.6 Previous Heart Problems -0.00-0.01-0.000.00-0.000.001 0.01 1.00 0.01-0.02-0.000.000.02-0.02 0.01 0.00 0.00 - 0.4 Stress Level -0.02-0.020.000.01 0.02-0.000.01-0.01-0.01-0.020.00 1.00-0.01-0.00-0.00-0.000.000.001-0.01-0.00 Sedentary Hours Per Day -0.02 0.02-0.01 0.00 0.00 0.02-0.000.01-0.010.00-0.001.00 0.00-0.000.01-0.010.00-0.01 BMI -0.000.02 0.01-0.00-0.010.01-0.010.01 0.00 0.02 0.01-0.00-0.000.01 1.00-0.010.01-0.010.00 - 0.2 Triglycerides -0.00-0.010.01 0.01-0.000.00 0.01 0.00-0.020.01-0.00-0.010.01-0.01 0.0 Previous Heart Problems Heart Rate Obesity Medication Use Triglycerides Cholesterol Smoking Alcohol Consumption Exercise Hours Per Week BMI Physical Activity Days Per Week Family History Stress Leve Sedentary Hours Per Day Income Sleep Hours Per Day Heart Attack Risk

```
class_counts = df["Heart Attack Risk"].value_counts()
plt.figure(figsize =(10,8))
class_counts.plot(kind = "bar", color = ["lightblue","lightcoral"])
plt.title("Heart Attack Risk")
plt.xlabel("HAR")
plt.ylabel("count")
plt.ylabel("count")
plt.xticks([0,1], labels = ["low risk (0)", "high risk (1)"], rotation = 0)
plt.show()
```



```
HAR
for column in df.columns:
    unique_values = df[column].unique()
    print(f"unique values for column {column}:\n{unique_values}\n")
unique values for column Patient ID:
['BMW7812' 'CZE1114' 'BNI9906' ... 'XKA5925' 'EPE6801' 'ZWN9666']
unique values for column Age:
[67 21 84 66 54 90 20 43 73 71 77 60 88 69 38 50 45 36 48 40 79 63 27 25
 86 42 52 29 30 47 44 33 51 70 85 31 56 24 74 72 55 26 53 46 57 22 35 39
 80 65 83 82 28 19 75 18 34 37 89 32 49 23 59 62 64 61 76 41 87 81 58 78
 68]
unique values for column Sex:
['Male' 'Female']
unique values for column Cholesterol:
[208 389 324 383 318 297 358 220 145 248 373 374 228 259 122 379 166 303
 340 294 359 202 133 159 271 273 328 154 135 197 321 375 360 263 201 347
 129 229 251 121 190 185 279 336 192 180 203 368 222 243 218 120 285 377
 369 311 139 266 153 339 329 333 398 124 183 163 362 390 200 396 255 209
 247 250 227 246 223 330 195 194 178 155 240 237 216 276 224 326 198 301
 314 304 334 213 254 230 316 277 388 206 384 205 261 308 338 382 291 168
```

```
171 378 253 245 226 281 123 173 231 234 268 306 186 293 161 380 239 149
 320 219 335 265 126 307 270 225 193 148 296 136 364 353 252 232 387 299
 357 214 370 345 351 344 152 150 131 272 302 337 170 356 274 188 125 138
376 181 184 275 394 128 217 399 283 289 284 327 262 212 350 385 162 141
 361 244 295 287 144 354 363 352 140 196 172 319 325 331 392 147 187 346
 286 151 300 165 343 366 317 386 158 157 242 241 365 257 348 175 298 269
 267 397 310 341 204 127 290 280 132 322 179 199 143 312 288 395 189 156
 238 381 391 355 210 400 260 235 167 256 249 207 130 134 137 305 236 315
292 323 146 258 332 372 142 309 177 367 371 211 282 342 264 176 160 233
313 164 349 221 191 174 393 278 215 169 182]
unique values for column Blood Pressure:
['158/88' '165/93' '174/99' ... '137/94' '94/76' '119/67']
unique values for column Heart Rate:
Γ 72
     98
         73
             93
                 48
                     84 107
                             68
                                 55
                                     97
                                         70
                                             85 102
                                                     40
                                                         56 104
                                                                71
                                                                    69
                             45
                                 50
                                         44 106
                                                        65 101
                                                                    43
 66
     81
         52 105
                 96
                     74
                         49
                                     46
                                                 83
                                                     86
                                                                 51
  79
     90
         94
             78
                 92
                     54 109
                             61
                                 64
                                     82 110
                                             42
                                                 63
                                                     41 100
                                                            76
                                                                 75
                                                                    58
  53
     60
        77
             47
                 59
                     57
                             67
                                     99 80
                                             95 108
                         87
                                 88
                                                     89
                                                         62 103
                                                                 91]
unique values for column Diabetes:
[0 1]
unique values for column Family History:
[0 1]
unique values for column Smoking:
[1 0]
unique values for column Obesity:
[0 1]
unique values for column Alcohol Consumption:
[0 1]
unique values for column Exercise Hours Per Week:
18.08174797]
unique values for column Diet:
['Average' 'Unhealthy' 'Healthy']
unique values for column Previous Heart Problems:
[0 1]
unique values for column Medication Use:
[0 1]
unique values for column Stress Level:
```

95 135 646 337 226 710 608 208 724 704 512 206 224 622 598 98 166 465 119 293 630 386 513 45 578 261 217 715 282 391 580 192 399 249 396 49 304 157 150 545 627 582 178 263 278 448 782 419 503 220 66 763 256 139 651 756 372 345 48 46 421 43 771 210 781 41 508 353 566 726 736 326 759 477 369 188 104 329 309 384 599 415 770 571 552 145 632 373 71 550 583 322 475 357 673 454 757 201 100 274 258 613 233 330 731 761 296 573 335 716 642 142 674 572 638 222 752 740 397 594 705 381 615 539 242 499 435 680 535 238 283 89 589 666 678 76 176 620 75 721 143 723 570 44 203 259 677 734 662 707 745 487 577 443 120 111 365 116 538 162 742 212 581 313 36 400 619 609 252 706 264 290 138 300 346 712 34 387 140 154 758 462 672 713 86 414 699 529 382 432 368 193 72 537 560 189 342 531 311 241 685 497 640 321 480 144 585 171 727 660 799 600

```
597 213 708 151 265 618 658 746 307 53 514 611 153 352 225 567 702 520
 417 102 607 548 647 476 762 147 424 459 409
                                             74 510 37 323 240 175 786
 080 439 504 772 670
                     59 334 703 392 90 496 422 279 343 671 794 163 328
 625 272 227 152 105 693 96 484 568 633 659 230 112 793 101 172 110 612
 185 289 418 533 686 641 169 349 173 516 62 557 596 728 371 738 444 561
 114 765 338 588 246 295 564 488 177 687 395 518 127 639 137 354 271 107
 340 534 768 130 720 405 430 268 108 748 351 393 361 170 470]
unique values for column Physical Activity Days Per Week:
[0 1 4 3 5 6 7 2]
unique values for column Sleep Hours Per Day:
[67451089]
unique values for column Country:
['Argentina' 'Canada' 'France' 'Thailand' 'Germany' 'Japan' 'Brazil'
 'South Africa' 'United States' 'Vietnam' 'China' 'Italy' 'Spain' 'India'
 'Nigeria' 'New Zealand' 'South Korea' 'Australia' 'Colombia'
 'United Kingdom']
unique values for column Continent:
['South America' 'North America' 'Europe' 'Asia' 'Africa' 'Australia']
unique values for column Hemisphere:
['Southern Hemisphere' 'Northern Hemisphere']
unique values for column Heart Attack Risk:
[0 1]
from sklearn.preprocessing import LabelEncoder
x=[]
lab=LabelEncoder()
for i in df.select_dtypes(include='object').columns.values:
    df[i]=lab.fit_transform(df[i])
for i in df.columns.values:
    print(df[i].value_counts())
    print()
521
        1
1038
        1
1601
7555
        1
4141
        1
       . .
4539
        1
1663
1538
        1
4056
        1
8719
        1
```

```
Name: Patient ID, Length: 8763, dtype: int64
90
      152
42
      150
33
      147
59
      147
29
      137
     . . .
75
      102
72
      101
39
      100
47
       99
51
       82
Name: Age, Length: 73, dtype: int64
     6111
1
     2652
0
Name: Sex, dtype: int64
235
       52
       47
360
149
       46
218
       46
251
       45
       . .
248
       20
186
       20
328
       20
398
       20
397
       19
Name: Cholesterol, Length: 281, dtype: int64
2005
        8
87
        8
        7
283
98
        7
        7
3295
2346
        1
2318
        1
3395
        1
3863
        1
838
        1
Name: Blood Pressure, Length: 3915, dtype: int64
94
       157
97
       146
57
       143
52
       140
```

```
104
       139
70
       107
48
       107
79
       105
        97
96
        93
73
Name: Heart Rate, Length: 71, dtype: int64
1
     5716
     3047
Name: Diabetes, dtype: int64
0
     4443
1
     4320
Name: Family History, dtype: int64
1
     7859
      904
0
Name: Smoking, dtype: int64
1
     4394
     4369
Name: Obesity, dtype: int64
1
     5241
0
     3522
Name: Alcohol Consumption, dtype: int64
             1
4.168189
18.477430
             1
11.883523
             1
19.353157
             1
19.365546
             1
9.884039
             1
12.644947
             1
1.089868
             1
10.500477
             1
18.081748
Name: Exercise Hours Per Week, Length: 8763, dtype: int64
1
     2960
0
     2912
2
     2891
Name: Diet, dtype: int64
0
     4418
1
     4345
```

```
Name: Previous Heart Problems, dtype: int64
0
     4396
1
     4367
Name: Medication Use, dtype: int64
2
      913
4
      910
7
      903
9
      887
8
      879
3
      868
1
      865
5
      860
6
      855
10
      823
Name: Stress Level, dtype: int64
6.615001
             1
0.772688
             1
0.723868
             1
10.125510
             1
2.054331
             1
11.921800
             1
0.087028
             1
9.198925
             1
3.383760
             1
9.005234
             1
Name: Sedentary Hours Per Day, Length: 8763, dtype: int64
225278
          4
194461
          3
195282
          3
220507
          2
139451
          2
         . .
44744
          1
85563
          1
          1
20443
258704
          1
247338
          1
Name: Income, Length: 8615, dtype: int64
31.251233
             1
39.385227
             1
36.280438
             1
18.218558
             1
23.885840
```

```
• •
28.358868
             1
22.539845
             1
34.721372
              1
18.881817
             1
32.914151
              1
Name: BMI, Length: 8763, dtype: int64
799
       25
507
       22
121
       22
593
       22
469
       22
120
        3
        3
213
185
        3
295
        3
        2
130
Name: Triglycerides, Length: 771, dtype: int64
3
     1143
1
     1121
2
     1109
7
     1095
5
     1079
4
     1077
6
     1074
0
     1065
Name: Physical Activity Days Per Week, dtype: int64
10
      1293
8
      1288
6
      1276
7
      1270
5
      1263
9
      1192
4
      1181
Name: Sleep Hours Per Day, dtype: int64
      477
7
0
      471
2
      462
17
      457
1
      449
12
      448
      446
6
3
      440
4
      436
```

```
11
      435
      433
10
9
      431
15
      430
      429
5
16
      428
13
      425
19
      425
18
      420
8
      412
14
      409
Name: Country, dtype: int64
1
     2543
3
     2241
5
     1362
2
      884
0
      873
4
      860
Name: Continent, dtype: int64
0
     5660
     3103
Name: Hemisphere, dtype: int64
0
     5624
1
     3139
Name: Heart Attack Risk, dtype: int64
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import accuracy_score
# Features (X) and target (y)
features = df.drop(columns=['Patient ID', 'Heart Attack Risk'])
target = df['Heart Attack Risk']
# Split the data into training and testing sets
X_train, X_test, y_train, y_test = train_test_split(features, target, test_size=0.2,
random state=42)
# Initialize and train the RandomForestClassifier
classifier = RandomForestClassifier(random_state=42)
classifier.fit(X_train, y_train)
# Make predictions on the test set
predictions = classifier.predict(X_test)
# Calculate accuracy
accuracy = accuracy_score(y_test, predictions)
print("Accuracy:", accuracy)
Accuracy: 0.6400456360524814
```

4. Documentation & Reporting

Documentation is crucial to capture the project's journey and make the insights accessible. We will create a comprehensive report encompassing the following sections.

4.1. Project Outline

This section will provide an overview of the project's objective and approach. It will outline the structure of the report and what readers can expect to find.

4.2. Business Insights

The project's ultimate goal is to derive actionable business insights. This section will explain how the analysis findings are translated into valuable recommendations for decision-making and strategy development.

4.3. Dataset Description

Detailed information on the selected datasets, including the reasons for their selection and how they were integrated into IBM Cloud Databases.

4.4. Database Setup and Data Exploration

A comprehensive explanation of the database setup and data exploration process. It will shed light on the infrastructure and methods employed to explore the datasets.

4.5. Analysis Techniques and Visualization

This section will delve into the analysis techniques used and the methods of visualization applied to convey the results effectively.

5. Conclusion:

In this part we will document our project and prepare it for submission. Document the big data analysis project and prepare it for submission.

Documentation:

Outline the project's objective, design thinking process, and development phases. Describe the selected dataset, database setup, analysis techniques, and visualization methods used. Explained how the analysis findings translate into valuable business insights.

Remember to tailor this structure to your specific project and organization's needs. Each section should provide a clear and concise overview of the project's details and its value to the business.