College Affordability

April 14, 2024

1 1. Prepare Datasets

1.1 1.1 Import the S3 data into SageMaker

```
[]: # Import packages
     import boto3
     import sagemaker
     import pandas as pd
     import numpy as np
     import matplotlib.pyplot as plt
     import seaborn as sns
     from sklearn.model_selection import train_test_split, RandomizedSearchCV
     from sklearn.linear_model import LogisticRegression
     from sklearn.preprocessing import StandardScaler
     from sklearn.pipeline import Pipeline
     from sklearn.metrics import accuracy_score, classification_report
     from scipy.stats import uniform
     from sagemaker.image_uris import retrieve
     from sagemaker.inputs import TrainingInput
    sagemaker.config INFO - Not applying SDK defaults from location:
    /etc/xdg/sagemaker/config.yaml
    sagemaker.config INFO - Not applying SDK defaults from location:
    /root/.config/sagemaker/config.yaml
[]: sess = sagemaker.Session()
    bucket = sess.default_bucket()
     role = sagemaker.get execution role()
     region = boto3.Session().region_name
     account_id = boto3.client("sts").get_caller_identity().get("Account")
     sm = boto3.Session().client(service_name="sagemaker", region_name=region)
    sagemaker.config INFO - Not applying SDK defaults from location:
    /etc/xdg/sagemaker/config.yaml
    sagemaker.config INFO - Not applying SDK defaults from location:
    /root/.config/sagemaker/config.yaml
    sagemaker.config INFO - Not applying SDK defaults from location:
```

```
/etc/xdg/sagemaker/config.yaml
sagemaker.config INFO - Not applying SDK defaults from location:
/root/.config/sagemaker/config.yaml
```

1.2 1.2 Store S3 locations

2024-03-31 20:06:23

```
[]: s3_public_path_kaggle = "s3://collegeaffordability317/Kaggle/"
     s3_public_path_tuition = "s3://collegeaffordability317/TuitionTracker/"
    s3_public_path_usda = "s3://collegeaffordability317/USDA/"
[]: %store s3_public_path_kaggle
    %store s3_public_path_tuition
    %store s3_public_path_usda
    Stored 's3_public_path_kaggle' (str)
    Stored 's3_public_path_tuition' (str)
    Stored 's3_public_path_usda' (str)
    1.3 Show all the data files for the project
```

```
[]: !aws s3 ls $s3_public_path_kaggle
    2024-03-25 00:16:36
                          14081604 college_data.csv
[]: !aws s3 ls $s3_public_path_tuition --recursive
                                 0 TuitionTracker/DataDictionary/
    2024-03-24 19:25:25
    2024-03-24 19:25:37
                             21527 TuitionTracker/DataDictionary/DataDictionary.xlsx
                                 0 TuitionTracker/GradRates/
    2024-03-24 19:24:22
    2024-03-24 19:24:45
                            838246 TuitionTracker/GradRates/GradRates.csv
    2024-03-24 19:24:55
                                 0 TuitionTracker/NetPrice/
    2024-03-24 19:25:12
                           1319314 TuitionTracker/NetPrice/NetPrice.csv
    2024-03-24 19:23:55
                                 0 TuitionTracker/RetentionRates/
    2024-03-24 19:24:08
                           1382164 TuitionTracker/RetentionRates/RetentionRates.csv
                                 0 TuitionTracker/StickerPrice/
    2024-03-24 19:20:50
    2024-03-24 19:21:49
                            653000 TuitionTracker/StickerPrice/StickerPrice.csv
[]: !aws s3 ls $s3_public_path_usda --recursive
    2024-03-24 19:26:33
                                 0 USDA/Education/
                           1610608 USDA/Education/Education.xlsx
    2024-03-24 19:26:58
    2024-03-24 19:28:03
                                 0 USDA/PovertyEstimates/
                            449799 USDA/PovertyEstimates/PovertyEstimates.csv
    2024-03-24 23:00:25
    2024-03-25 22:15:58
                            187092 USDA/PovertyTableWithState/20240325_221556_00069_
    795ia_cfa99626-d392-4100-8b4f-0cfe8f1c1388.gz
    2024-03-25 23:21:35
                            187092 USDA/PovertyTableWithState/20240325_232134_00063_
    r69ka_db087c21-7823-41ba-8bf6-04d122f8c50d.gz
```

187092 USDA/PovertyTableWithState/20240331_200621_00076_

2 2. Athena DB

2.1 2.1 Data Wrangling

```
[]: from pyathena import connect
     # Set S3 staging directory -- this is a temporary directory used for Athenau
      \hookrightarrow queries
     s3_staging_dir = "s3://{0}/athena/staging".format(bucket)
[]: # Set Athena parameters
     college_affordability_database = 'collegeaffordability317'
     table_name = 'college_data'
[]: conn = connect(region_name=region, s3_staging_dir=s3_staging_dir)
[]: statement = "CREATE DATABASE IF NOT EXISTS {}".
      →format(college_affordability_database)
     pd.read sql(statement, conn)
[]: Empty DataFrame
     Columns: []
     Index: []
[]: statement = "SHOW DATABASES"
     df_show = pd.read_sql(statement, conn)
     df show.head(10)
[]:
                  database name
     0 collegeaffordability317
                        default
     1
     2
                         dsoaws
     3
         sagemaker_featurestore
         2.2 Populate Tables
    2.2
[]: tuition_tracker_dir = 's3://collegeaffordability317/TuitionTracker/'
     usda_dir = 's3://collegeaffordability317/USDA/'
     kaggle_dir = 's3://collegeaffordability317/Kaggle/'
```

```
[]: # Drop the table if it already exists
     university_table = 'University'
     pd.read_sql(f'DROP TABLE IF EXISTS {college_affordability_database}.

√{university_table}', conn)
     # Define the CREATE TABLE statement with data types in lowercase
     create_university_table = f"""
     CREATE EXTERNAL TABLE IF NOT EXISTS {college_affordability_database}.

√{university_table} (
         UNITID INT,
         INSTNM STRING,
         CITY STRING,
         STABBR STRING,
         ZIP STRING,
         REGION STRING,
         PREDDEG STRING,
         LOCALE INT,
         LATITUDE FLOAT,
         LONGITUDE FLOAT,
         CCBASIC STRING,
         CCUGPROF STRING,
         CCSIZSET STRING,
         ADM_RATE_ALL FLOAT,
         PPTUG_EF FLOAT,
         SAT_AVG_ALL INT,
         COSTT4_A INT,
         CONTROL STRING,
         TUITIONFEE IN INT,
         TUITIONFEE_OUT INT,
         MEDIAN_HH_INC FLOAT,
         MN_EARN_WNE_INDEPO_P10 INT,
         MN_EARN_WNE_INDEP1_P10 INT,
         UGDS_WHITE FLOAT,
         UGDS BLACK FLOAT,
         UGDS_HISP FLOAT,
         UGDS_ASIAN FLOAT,
         UGDS_AIAN FLOAT,
         UGDS_NHPI FLOAT,
         UGDS_2MOR FLOAT,
         UGDS_NRA FLOAT,
         UGDS_UNKN FLOAT,
         year INT
         ROW FORMAT DELIMITED
         FIELDS TERMINATED BY ','
         LOCATION '{kaggle_dir}'
         TBLPROPERTIES ('skip.header.line.count'='1')
```

```
0.00
     # Execute create table statement
     pd.read_sql(create_university_table, conn)
     pd.read_sql(f'SELECT * FROM {college_affordability_database}.{university_table}_
      []:
        unitid
                                                            city stabbr
                                             instnm
                                                                                zip
     0 100654
                           Alabama A & M University
                                                          Normal
                                                                              35762
                                                                     AL
               University of Alabama at Birmingham
     1 100663
                                                     Birmingham
                                                                     AL
                                                                         35294-0110
     2 100706
                University of Alabama in Huntsville
                                                     Huntsville
                                                                     ΑL
                                                                              35899
     3 100751
                          The University of Alabama
                                                     Tuscaloosa
                                                                     AL
                                                                         35487-0100
     4 100858
                                  Auburn University
                                                          Auburn
                                                                     AT.
                                                                              36849
                                                 region \
     O Southeast (AL AR FL GA KY LA MS NC SC TN VA WV)
     1 Southeast (AL AR FL GA KY LA MS NC SC TN VA WV)
     2 Southeast (AL AR FL GA KY LA MS NC SC TN VA WV)
     3 Southeast (AL AR FL GA KY LA MS NC SC TN VA WV)
     4 Southeast (AL AR FL GA KY LA MS NC SC TN VA WV)
                                         preddeg locale latitude longitude
     O Predominantly bachelor's-degree granting
                                                   None
                                                             None
                                                                       None
     1 Predominantly bachelor's-degree granting
                                                   None
                                                             None
                                                                       None ...
     2 Predominantly bachelor's-degree granting
                                                   None
                                                             None
                                                                       None
     3 Predominantly bachelor's-degree granting
                                                   None
                                                             None
                                                                       None
     4 Predominantly bachelor's-degree granting
                                                                       None ...
                                                   None
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       ugds_white ugds_black ugds_hisp ugds_asian
                                                   ugds_aian ugds_nhpi ugds_2mor
     0
             None
                        None
                                  None
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                                                                     None
                                                                               None
             None
     1
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                                  None
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     2
             None
                        None
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     3
             None
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             None
                        None
                                  None
                                              None
                                                          None
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                                                                               None
       ugds_nra
                ugds_unkn
                            year
     0
         0.0402
                    0.0017
                            2001
         0.0330
                    0.0255
                            2001
     1
     2
         0.0396
                    0.0000
                            2001
         0.0159
                    0.0000
                            2001
         0.0084
                    0.0016
                            2001
     [5 rows x 33 columns]
[]: # Drop the table if it already exists
     sticker_price_table = 'StickerPrice'
```

```
pd.read_sql(f'DROP TABLE IF EXISTS {college_affordability_database}.
 # Define the CREATE TABLE statement with data types in lowercase
create_sticker_price_table = f"""
CREATE EXTERNAL TABLE IF NOT EXISTS {college affordability database}.
 unit_id INT,
   institution_name STRING,
    sector INT,
   total_price_in_state_on_campus_2021_2022 FLOAT,
   total price in state off campus wo fam 2021 2022 FLOAT,
   total_price_in_state_off_campus_w_fam_2021_2022 FLOAT,
   total_price_in_state_on_campus_2020_2021 FLOAT,
   total_price_in_state_off_campus_wo_fam_2020_2021 FLOAT,
   total_price_in_state_off_campus_w_fam_2020_2021 FLOAT,
   total_price_in_state_on_campus_2019_2020 FLOAT,
   total_price_in_state_off_campus_wo_fam_2019_2020 FLOAT,
   total_price_in_state_off_campus_w_fam_2019_2020 FLOAT,
   total_price_in_state_on_campus_2018_2019 FLOAT,
   total_price_in_state_off_campus_wo_fam_2018_2019 FLOAT,
   total_price_in_state_off_campus_w_fam_2018_2019 FLOAT,
   total_price_in_state_on_campus_2017_2018 FLOAT,
   total_price_in_state_off_campus_wo_fam_2017_2018 FLOAT,
   total_price_in_state_off_campus_w_fam_2017_2018 FLOAT,
   total_price_in_state_on_campus_2016_2017 FLOAT,
   total price in state off campus wo fam 2016 2017 FLOAT,
   total_price_in_state_off_campus_w_fam_2016_2017 FLOAT,
   total_price_in_state_on_campus_2015_2016 FLOAT,
   total_price_in_state_off_campus_wo_fam_2015_2016 FLOAT,
   total_price_in_state_off_campus_w_fam_2015_2016 FLOAT,
   total_price_in_state_on_campus_2014_2015 FLOAT,
   total_price_in_state_off_campus_wo_fam_2014_2015 FLOAT,
   total_price_in_state_off_campus_w_fam_2014_2015 FLOAT,
   total_price_in_state_on_campus_2013_2014 FLOAT,
   total_price_in_state_off_campus_wo_fam_2013_2014 FLOAT,
   total_price_in_state_off_campus_w_fam_2013_2014 FLOAT,
   total_price_in_state_on_campus_2012_2013 FLOAT,
   total_price_in_state_off_campus_wo_fam_2012_2013 FLOAT,
   total_price_in_state_off_campus_w_fam_2012_2013 FLOAT,
   total_price_in_state_on_campus_2011_2012 FLOAT,
   total_price_in_state_off_campus_wo_fam_2011_2012 FLOAT,
   total_price_in_state_off_campus_w_fam_2011_2012 FLOAT
   ROW FORMAT DELIMITED
   FIELDS TERMINATED BY ','
   LOCATION '{tuition_tracker_dir}/{sticker_price_table}'
```

```
TBLPROPERTIES ('skip.header.line.count'='1')
    0.00
    # Execute create table statement
    pd.read_sql(create_sticker_price_table, conn)
    pd.read_sql(f'SELECT * FROM {college_affordability_database}.
      []:
       unit_id
                                                 institution_name sector
        180203
    0
                                           Aaniiih Nakoda College
    1
        222178
                                     Abilene Christian University
                                                                        2
        497037
                Abilene Christian University-Undergraduate Online
                                                                        2
        138558
                             Abraham Baldwin Agricultural College
    3
                                                                         1
        488031
                                       Abraham Lincoln University
                                                                         3
       total_price_in_state_on_campus_2021_2022
    0
    1
                                         55500.0
    2
                                            NaN
    3
                                         15727.0
    4
                                            NaN
       total_price_in_state_off_campus_wo_fam_2021_2022
    0
                                                 17030.0
    1
                                                55500.0
    2
                                                30670.0
    3
                                                 13965.0
    4
                                                27133.0
       total_price_in_state_off_campus_w_fam_2021_2022
    0
                                                8510.0
    1
                                                43872.0
    2
                                                19042.0
    3
                                                7765.0
    4
                                                11365.0
       total_price_in_state_on_campus_2020_2021
    0
                                            NaN
                                         53672.0
    1
    2
                                            NaN
    3
                                         15575.0
    4
                                            NaN
       total_price_in_state_off_campus_wo_fam_2020_2021
    0
                                                 17030.0
    1
                                                53672.0
```

```
2
                                                  NaN
3
                                              13865.0
4
                                              25576.0
   total_price_in_state_off_campus_w_fam_2020_2021
0
                                              8510.0
                                             42322.0
1
2
                                                 NaN
3
                                              7665.0
4
                                             11176.0
   total_price_in_state_on_campus_2019_2020
0
                                          NaN
1
                                      51887.0
2
                                          NaN
3
                                      15479.0
4
                                          NaN
   total_price_in_state_off_campus_w_fam_2014_2015
0
                                              8510.0
                                             34100.0
1
2
                                                 NaN
3
                                              6894.0
                                                 NaN
   total_price_in_state_on_campus_2013_2014
0
                                          NaN
1
                                      41800.0
2
                                          NaN
3
                                      17503.0
4
                                          NaN
   total_price_in_state_off_campus_wo_fam_2013_2014
0
                                              17030.0
                                              41800.0
1
2
                                                  NaN
3
                                              13188.0
                                                  NaN
   total_price_in_state_off_campus_w_fam_2013_2014
0
                                              8510.0
                                             33000.0
1
2
                                                 NaN
3
                                              7578.0
4
                                                 NaN
   total_price_in_state_on_campus_2012_2013 \
```

```
0
                                               NaN
     1
                                           39900.0
     2
                                               NaN
     3
                                           16550.0
     4
                                               NaN
        total_price_in_state_off_campus_wo_fam_2012_2013
    0
                                                   17030.0
     1
                                                   39900.0
     2
                                                       NaN
     3
                                                   12619.0
     4
                                                       NaN
        total_price_in_state_off_campus_w_fam_2012_2013 \
     0
                                                   8510.0
                                                  31250.0
     1
     2
                                                      NaN
     3
                                                   7009.0
     4
                                                      NaN
        total_price_in_state_on_campus_2011_2012
    0
                                               NaN
     1
                                           38250.0
     2
                                               NaN
     3
                                           12347.0
     4
                                               NaN
        total_price_in_state_off_campus_wo_fam_2011_2012 \
     0
                                                   17030.0
     1
                                                       NaN
     2
                                                       NaN
     3
                                                       NaN
     4
                                                       NaN
        total_price_in_state_off_campus_w_fam_2011_2012
     0
                                                   8510.0
     1
                                                      NaN
     2
                                                      NaN
     3
                                                      NaN
                                                      NaN
     [5 rows x 36 columns]
[]: # Drop the table if it already exists
    net_price_table = 'NetPrice'
    pd.read_sql(f'DROP TABLE IF EXISTS {college_affordability_database}.

¬{net_price_table}', conn)
```

```
# Define the CREATE TABLE statement with data types in lowercase
create_net_price_table = f"""
CREATE EXTERNAL TABLE IF NOT EXISTS {college_affordability_database}.
 →{net_price_table} (
   unit id int,
    institution_name string,
    sector int,
   avg_net_price_grants_scholarship_2020_2021 float,
   avg net price income 0 30k titleiv fed finaid 2020 2021 float,
   avg_net_price_income_30k_48k_titleiv_fed_finaid_2020_2021 float,
   avg_net_price_income_48k_75k_titleiv_fed_finaid_2020_2021 float,
   avg net price income 75k 110k titleiv fed finaid 2020 2021 float,
   avg_net_price_income_over_110k_titleiv_fed_finaid_2020_2021 float,
   avg_net_price_grants_scholarship_2019_2020 float,
   avg_net_price_income_0_30k_titleiv_fed_finaid_2019_2020 float,
   avg_net_price_income_30k_48k_titleiv_fed_finaid_2019_2020 float,
   avg_net_price_income_48k_75k_titleiv_fed_finaid_2019_2020 float,
   avg_net_price_income_75k_110k_titleiv_fed_finaid_2019_2020 float,
   avg_net_price_income_over_110k_titleiv_fed_finaid_2019_2020 float,
   avg net price grants scholarship 2018 2019 float,
   avg_net_price_income_0_30k_titleiv_fed_finaid_2018_2019 float,
   avg_net_price_income_30k_48k_titleiv_fed_finaid_2018_2019 float,
   avg_net_price_income_48k_75k_titleiv_fed_finaid_2018_2019 float,
   avg_net_price_income_75k_110k_titleiv_fed_finaid_2018_2019 float,
   avg_net_price_income_over_110k_titleiv_fed_finaid_2018_2019 float,
   avg_net_price_grants_scholarship_2017_2018 float,
   avg_net_price_income_0_30k_titleiv_fed_finaid_2017_2018_float,
   avg_net_price_income_30k_48k_titleiv_fed_finaid_2017_2018 float,
   avg_net_price_income_48k_75k_titleiv_fed_finaid_2017_2018 float,
   avg_net_price_income_75k_110k_titleiv_fed_finaid_2017_2018 float,
   avg_net_price_income_over_110k_titleiv_fed_finaid_2017_2018 float,
   avg_net_price_grants_scholarship_2016_2017 float,
   avg_net_price_income_0_30k_titleiv_fed_finaid_2016_2017 float,
   avg_net_price_income_30k_48k_titleiv_fed_finaid_2016_2017 float,
   avg_net_price_income_48k_75k_titleiv_fed_finaid_2016_2017 float,
   avg_net_price_income_75k_110k_titleiv_fed_finaid_2016_2017 float,
   avg_net_price_income_over_110k_titleiv_fed_finaid_2016_2017 float,
   avg_net_price_grants_scholarship_2015_2016 float,
   avg_net_price_income_0_30k_titleiv_fed_finaid_2015_2016 float,
   avg_net_price_income_30k_48k_titleiv_fed_finaid_2015_2016 float,
   avg_net_price_income_48k_75k_titleiv_fed_finaid_2015_2016 float,
   avg_net_price_income_75k_110k_titleiv_fed_finaid_2015_2016_float,
   avg_net_price_income_over_110k_titleiv_fed_finaid_2015_2016 float,
   avg_net_price_grants_scholarship_2014_2015 float,
   avg_net_price_income_0_30k_titleiv_fed_finaid_2014_2015 float,
    avg_net_price_income_30k_48k_titleiv_fed_finaid_2014_2015 float,
```

```
avg net price income 48k 75k titleiv fed finaid 2014 2015 float,
         avg_net_price_income_75k_110k_titleiv_fed_finaid_2014_2015 float,
         avg net price income over 110k titleiv fed finaid 2014 2015 float,
         avg_net_price_grants_scholarship_2013_2014 float,
         avg_net_price_income_0_30k_titleiv_fed_finaid_2013_2014 float,
         avg_net_price_income_30k_48k_titleiv_fed_finaid_2013_2014 float,
         avg_net_price_income_48k_75k_titleiv_fed_finaid_2013_2014 float,
         avg_net_price_income_75k_110k_titleiv_fed_finaid_2013_2014 float,
         avg net price income over 110k titleiv fed finaid 2013 2014 float,
         avg_net_price_grants_scholarship_2012_2013 float,
         avg_net_price_income_0_30k_titleiv_fed_finaid_2012_2013 float,
         avg_net_price_income_30k_48k_titleiv_fed_finaid_2012_2013 float,
         avg_net_price_income_48k_75k_titleiv_fed_finaid_2012_2013 float,
         avg net price income 75k 110k titleiv fed finaid 2012 2013 float,
         avg_net_price_income_over_110k_titleiv_fed_finaid_2012_2013 float,
         avg_net_price_grants_scholarship_2011_2012 float,
         avg_net_price_income_0_30k_titleiv_fed_finaid_2011_2012 float,
         avg net price income 30k 48k titleiv fed finaid 2011 2012 float,
         avg_net_price_income_48k_75k_titleiv_fed_finaid_2011_2012 float,
         avg_net_price_income_75k_110k_titleiv_fed_finaid_2011_2012 float,
         avg_net_price_income_over_110k_titleiv_fed_finaid_2011_2012 float
         )
         ROW FORMAT DELIMITED
         FIELDS TERMINATED BY '.'
         LOCATION '{tuition_tracker_dir}/{net_price_table}'
         TBLPROPERTIES ('skip.header.line.count'='1')
     # Execute create table statement
     pd.read_sql(create_net_price_table, conn)
     pd.read_sql(f'SELECT * FROM {college_affordability_database}.{net_price_table}_\( \)
      →LIMIT 5', conn)
[]:
       unit_id
                                                  institution_name sector \
     0
         180203
                                            Aaniiih Nakoda College
                                                                          1
     1
         222178
                                      Abilene Christian University
                                                                          2
     2
        497037 Abilene Christian University-Undergraduate Online
                                                                          2
     3
         138558
                              Abraham Baldwin Agricultural College
                                                                          1
     4
        488031
                                        Abraham Lincoln University
                                                                          3
       avg_net_price_grants_scholarship_2020_2021 \
     0
                                            8381.0
     1
                                               NaN
     2
                                               NaN
     3
                                            7744.0
     4
                                               NaN
```

```
avg_net_price_income_0_30k_titleiv_fed_finaid_2020_2021 \
0
                                                8119.0
1
                                                   NaN
2
                                                   NaN
3
                                                4784.0
4
                                                   NaN
   avg_net_price_income_30k_48k_titleiv_fed_finaid_2020_2021 \
0
                                                8326.0
1
                                                   NaN
2
                                                   NaN
3
                                                5862.0
4
                                                   NaN
   avg_net_price_income_48k_75k_titleiv_fed_finaid_2020_2021 \
0
                                               10138.0
1
                                                   NaN
2
                                                   NaN
3
                                                8408.0
4
                                                   NaN
   avg_net_price_income_75k_110k_titleiv_fed_finaid_2020_2021 \
0
                                                   NaN
1
                                                   NaN
2
                                                   NaN
3
                                               10953.0
4
                                                   NaN
   avg_net_price_income_over_110k_titleiv_fed_finaid_2020_2021 \
0
                                                   NaN
1
                                                   NaN
2
                                                   NaN
3
                                               10568.0
                                                   NaN
   avg_net_price_grants_scholarship_2019_2020
0
                                        7777.0
1
                                           NaN
2
                                            NaN
3
                                        8106.0
4
                                            NaN
   avg_net_price_income_30k_48k_titleiv_fed_finaid_2012_2013 \
0
                                                5024.0
1
                                                   NaN
2
                                                   NaN
```

```
3
                                                8862.0
4
                                                   NaN
   avg_net_price_income_48k_75k_titleiv_fed_finaid_2012_2013 \
0
                                                3359.0
1
                                                   NaN
2
                                                   NaN
3
                                               10959.0
4
                                                   NaN
   avg_net_price_income_75k_110k_titleiv_fed_finaid_2012_2013 \
0
                                                   NaN
1
2
                                                   NaN
3
                                               12342.0
4
                                                   NaN
   avg_net_price_income_over_110k_titleiv_fed_finaid_2012_2013 \
0
                                                   NaN
                                                   NaN
1
2
                                                   NaN
                                               12946.0
3
4
                                                   NaN
   avg_net_price_grants_scholarship_2011_2012 \
0
                                       13201.0
1
                                            NaN
2
                                           NaN
3
                                        7518.0
4
                                           NaN
   avg_net_price_income_0_30k_titleiv_fed_finaid_2011_2012 \
0
                                               13133.0
1
                                                   NaN
2
                                                   NaN
3
                                                6026.0
4
                                                   NaN
   avg_net_price_income_30k_48k_titleiv_fed_finaid_2011_2012 \
                                               13769.0
0
1
                                                   NaN
2
                                                   NaN
                                                6895.0
3
4
                                                   NaN
   avg_net_price_income_48k_75k_titleiv_fed_finaid_2011_2012 \
0
                                               14069.0
```

```
2
                                                       NaN
     3
                                                    9511.0
     4
                                                       NaN
        avg_net_price_income_75k_110k_titleiv_fed_finaid_2011_2012 \
     0
                                                       NaN
     1
                                                       NaN
     2
                                                       NaN
     3
                                                   11080.0
     4
                                                       NaN
        avg_net_price_income_over_110k_titleiv_fed_finaid_2011_2012
     0
                                                       NaN
     1
                                                       NaN
     2
                                                       NaN
     3
                                                   11182.0
     4
                                                       NaN
     [5 rows x 63 columns]
[]: # Drop the table if it already exists
     poverty_table = 'PovertyEstimates'
     pd.read_sql(f'DROP TABLE IF EXISTS {college_affordability_database}.
      →{poverty_table}', conn)
     # Define the CREATE TABLE statement with data types in lowercase
     create_poverty_table = f"""
     CREATE EXTERNAL TABLE IF NOT EXISTS {college_affordability_database}.
      →{poverty_table} (
         FIPS_Code INT,
         Stabr STRING,
         Area_name STRING,
         Rural urban Continuum Code 2003 STRING,
         Urban Influence Code 2003 STRING,
         Rural urban Continuum Code 2013 STRING,
         Urban_Influence_Code_2013 STRING,
         POVALL_2021 STRING,
         CI90LBALL_2021 STRING,
         CI90UBALL_2021 STRING,
         PCTPOVALL_2021 STRING,
         CI90LBALLP_2021 STRING,
         CI90UBALLP_2021 STRING,
         POV017_2021 STRING,
         CI90LB017_2021 STRING,
         CI90UB017_2021 STRING,
         PCTPOV017_2021 STRING,
```

NaN

1

```
CI90LB017P_2021 STRING,
         CI90UB017P_2021 STRING,
         POV517_2021 STRING,
         CI90LB517_2021 STRING,
         CI90UB517_2021 STRING,
         PCTPOV517_2021 STRING,
         CI90LB517P_2021 STRING,
         CI90UB517P_2021 STRING,
         MEDHHINC 2021 STRING,
         CI90LBINC 2021 STRING,
         CI90UBINC_2021 STRING,
         POV04_2021 STRING,
         CI90LB04_2021 STRING,
         CI90UB04_2021 STRING,
         PCTPOVO4_2021 STRING,
         CI90LB04P_2021 STRING,
         CI90UB04P_2021 STRING
         ROW FORMAT DELIMITED
         FIELDS TERMINATED BY ','
         LOCATION '{usda_dir}/{poverty_table}'
         TBLPROPERTIES ('skip.header.line.count'='1')
     0.00
     # Execute create table statement
     pd.read_sql(create_poverty_table, conn)
     pd.read_sql(f'SELECT * FROM {college_affordability_database}.{poverty_table}_u
      →LIMIT 5', conn)
[]:
        fips_code stabr
                              area_name rural_urban_continuum_code_2003 \
     0
                          United States
                0
                     US
     1
             1000
                     AL
                                Alabama
                                                                       2
             1001
                     AL Autauga County
     3
             1003
                     AL Baldwin County
                                                                       4
     4
             1005
                     AL Barbour County
      urban_influence_code_2003 rural_urban_continuum_code_2013 \
     0
     1
     2
                               2
                                                                2
     3
                               5
                                                                3
      urban_influence_code_2013 povall_2021 ci90lball_2021 ci90uball_2021 ... \
     0
                                                    41149497
                                    41393176
                                                                   41636855 ...
     1
                                      800848
                                                      782169
                                                                     819527 ...
```

```
2
                               2
                                        6296
                                                       4772
                                                                      7820 ...
     3
                               2
                                       25526
                                                      21599
                                                                     29453 ...
     4
                               6
                                        5089
                                                       3773
                                                                      6405 ...
       ci90ub517p_2021 medhhinc_2021 ci90lbinc_2021 ci90ubinc_2021 pov04_2021
     0
                  16.3
                               69717
                                              69583
                                                             69851
                                                                      3349149
                  22.5
                               53990
                                              53218
                                                             54762
                                                                        71220
     1
     2
                  20.4
                               66444
                                              60061
                                                             72827
     3
                                              60723
                  18.5
                               65658
                                                             70593
     4
                  44.6
                                              34308
                               38649
                                                             42990
       ci901b04_2021 ci90ub04_2021 pctpov04_2021 ci901b04p_2021 ci90ub04p_2021
     0
             3299669
                           3398629
                                            18.3
                                                             18
     1
               66888
                             75552
                                            25.1
                                                           23.6
                                                                           26.6
     2
     3
     4
     [5 rows x 34 columns]
[]: # Drop the table if it already exists
     unemployment_table = 'Unemployment'
     pd.read_sql(f'DROP TABLE IF EXISTS {college_affordability_database}.

√{unemployment_table}', conn)
     # Define the CREATE TABLE statement with data types in lowercase
     create_unemployment_table = f"""
     CREATE EXTERNAL TABLE IF NOT EXISTS {college affordability database}.
      FIPS_Code INT,
         State STRING,
         Area Name STRING,
         Rural_Urban_Continuum_Code_2013 INT,
         Urban Influence Code 2013 INT,
         Metro_2013 INT,
         Civilian labor force 2000 INT,
         Employed_2000 INT,
         Unemployed_2000 INT,
         Unemployment_rate_2000 FLOAT,
         Civilian_labor_force_2001 INT,
         Employed_2001 INT,
         Unemployed_2001 INT,
         Unemployment_rate_2001 FLOAT,
         Civilian_labor_force_2002 INT,
         Employed_2002 INT,
         Unemployed_2002 INT,
         Unemployment_rate_2002 FLOAT,
```

Civilian_labor_force_2003 INT, Employed_2003 INT, Unemployed_2003 INT, Unemployment_rate_2003 FLOAT, Civilian_labor_force_2004 INT, Employed_2004 INT, Unemployed_2004 INT, Unemployment_rate_2004 FLOAT, Civilian labor force 2005 INT, Employed_2005 INT, Unemployed_2005 INT, Unemployment_rate_2005 FLOAT, Civilian_labor_force_2006 INT, Employed_2006 INT, Unemployed_2006 INT, Unemployment_rate_2006 FLOAT, Civilian_labor_force_2007 INT, Employed_2007 INT, Unemployed_2007 INT, Unemployment_rate_2007 FLOAT, Civilian_labor_force_2008 INT, Employed 2008 INT, Unemployed_2008 INT, Unemployment rate 2008 FLOAT, Civilian_labor_force_2009 INT, Employed 2009 INT, Unemployed_2009 INT, Unemployment_rate_2009 FLOAT, Civilian_labor_force_2010 INT, Employed_2010 INT, Unemployed_2010 INT, Unemployment_rate_2010 FLOAT, Civilian_labor_force_2011 INT, Employed_2011 INT, Unemployed_2011 INT, Unemployment_rate_2011 FLOAT, Civilian labor force 2012 INT, Employed_2012 INT, Unemployed 2012 INT, Unemployment_rate_2012 FLOAT, Civilian labor force 2013 INT, Employed_2013 INT, Unemployed_2013 INT, Unemployment_rate_2013 FLOAT, Civilian_labor_force_2014 INT, Employed_2014 INT, Unemployed_2014 INT,

```
Unemployment_rate_2014 FLOAT,
   Civilian_labor_force_2015 INT,
    Employed_2015 INT,
   Unemployed_2015 INT,
   Unemployment_rate_2015 FLOAT,
   Civilian_labor_force_2016 INT,
   Employed_2016 INT,
   Unemployed_2016 INT,
   Unemployment rate 2016 FLOAT,
   Civilian_labor_force_2017 INT,
   Employed_2017 INT,
   Unemployed_2017 INT,
   Unemployment_rate_2017 FLOAT,
   Civilian_labor_force_2018 INT,
   Employed_2018 INT,
   Unemployed_2018 INT,
   Unemployment_rate_2018 FLOAT,
   Civilian_labor_force_2019 INT,
   Employed_2019 INT,
   Unemployed_2019 INT,
   Unemployment_rate_2019 FLOAT,
   Civilian_labor_force_2020 INT,
   Employed_2020 INT,
   Unemployed 2020 INT,
   Unemployment_rate_2020 FLOAT,
   Civilian_labor_force_2021 INT,
   Employed_2021 INT,
   Unemployed_2021 INT,
   Unemployment_rate_2021 FLOAT,
   Civilian_labor_force_2022 INT,
   Employed_2022 INT,
   Unemployed_2022 INT,
   Unemployment_rate_2022 FLOAT,
   Median_Household_Income_2021 INT,
   Med_HH_Income_Percent_of_State_Total_2021 FLOAT
   ROW FORMAT DELIMITED
   FIELDS TERMINATED BY ','
   LOCATION '{usda dir}/{unemployment table}'
   TBLPROPERTIES ('skip.header.line.count'='1')
# Execute create table statement
pd.read_sql(create_unemployment_table, conn)
pd.read_sql(f'SELECT * FROM {college_affordability_database}.
```

```
[]:
        fips_code state
                                              rural_urban_continuum_code_2013
                                   area_name
     0
                 0
                      US
                               United States
                                                                             NaN
             1000
     1
                      AT.
                                     Alabama
                                                                             NaN
     2
             1001
                      ΑL
                          Autauga County AL
                                                                             2.0
                          Baldwin County AL
     3
             1003
                                                                             3.0
     4
             1005
                          Barbour County AL
                                                                             6.0
        urban_influence_code_2013 metro_2013
                                                civilian_labor_force_2000 \
     0
                                NaN
                                            NaN
                                                                   142601576
                                NaN
                                            NaN
                                                                     2147173
     1
     2
                                             1.0
                                2.0
                                                                       21861
     3
                                2.0
                                             1.0
                                                                       69979
     4
                                6.0
                                             0.0
                                                                       11449
                        unemployed_2000
                                         unemployment_rate_2000
        employed_2000
            136904853
                                 5696723
     0
     1
               2047731
                                   99442
                                                               4.6
     2
                                     890
                 20971
                                                               4.1 ...
     3
                 67370
                                    2609
                                                               3.7
     4
                                     637
                 10812
                                                               5.6 ...
        civilian_labor_force_2021
                                    employed 2021 unemployed 2021
                         162229903
                                         153544980
                                                              8684923
     0
     1
                           2259349
                                            2183330
                                                                76019
     2
                              26545
                                              25809
                                                                  736
     3
                              99953
                                              97034
                                                                 2919
     4
                                                                  459
                               8280
                                               7821
        unemployment_rate_2021
                                 civilian_labor_force_2022
                                                               employed_2022
     0
                            5.4
                                                   164781642
                                                                   158766998
                            3.4
     1
                                                     2286028
                                                                     2226670
     2
                            2.8
                                                       26789
                                                                       26181
     3
                            2.9
                                                      102849
                                                                      100432
     4
                            5.5
                                                        8241
                                                                        7906
                          unemployment_rate_2022 median_household_income_2021
        unemployed_2022
     0
                 6014644
                                               3.7
                                                                             69717
                                               2.6
     1
                   59358
                                                                             53990
     2
                     608
                                               2.3
                                                                             66444
     3
                    2417
                                               2.4
                                                                             65658
     4
                     335
                                               4.1
                                                                             38649
        med_hh_income_percent_of_state_total_2021
     0
                                                 NaN
     1
                                               100.0
     2
                                               123.1
     3
                                               121.6
```

4 71.6

[5 rows x 100 columns]

2.3 2.3 Transform Tables

```
[]: # Load Athena tables into Pandas dataframes
     University = pd.read_sql(f'SELECT * FROM {college_affordability_database}.
      →{university_table}', conn)
     NetPrice = pd.read_sql(f'SELECT * FROM {college_affordability_database}.
      →{net_price_table}', conn)
     StickerPrice = pd.read_sql(f'SELECT * FROM {college_affordability_database}.

¬{sticker_price_table}', conn)
     Poverty = pd.read_sql(f'SELECT * FROM {college_affordability_database}.
      →{poverty_table}', conn)
     Unemployment = pd.read_sql(f'SELECT * FROM {college_affordability_database}.
      →{unemployment_table}', conn)
[]: # Combine the area name and state arrbriviation columns so we can join the table
     Poverty['area_name_with_state'] = Poverty['area_name'] + ' ' + Poverty['stabr']
     Poverty.head()
                              area_name rural_urban_continuum_code_2003 \
[]:
        fips_code stabr
                          United States
                0
                     US
     1
             1000
                     ΑL
                                Alabama
     2
                     AL Autauga County
                                                                       2
             1001
                     AL Baldwin County
             1003
                                                                       4
     4
             1005
                     AL Barbour County
                                                                       6
      urban_influence_code_2003 rural_urban_continuum_code_2013 \
     0
     1
                               2
                                                                2
     2
                               5
     3
                                                                3
      urban_influence_code_2013_povall_2021_ci90lball_2021_ci90uball_2021 ...
     0
                                    41393176
                                                    41149497
                                                                   41636855
     1
                                      800848
                                                                     819527 ...
                                                      782169
     2
                               2
                                                                       7820 ...
                                        6296
                                                        4772
                               2
     3
                                                       21599
                                                                      29453 ...
                                        25526
                               6
     4
                                        5089
                                                        3773
                                                                       6405 ...
      medhhinc_2021 ci90lbinc_2021 ci90ubinc_2021 pov04_2021 ci90lb04_2021 \
     0
               69717
                              69583
                                              69851
                                                       3349149
                                                                     3299669
```

```
2
                              60061
                                              72827
               66444
     3
               65658
                              60723
                                             70593
     4
               38649
                              34308
                                              42990
       ci90ub04_2021 pctpov04_2021 ci90lb04p_2021 ci90ub04p_2021 \
     0
             3398629
                              18.3
                                                18
               75552
                              25.1
                                             23.6
                                                             26.6
     1
     2
     3
     4
       area_name_with_state
     0
           United States US
     1
                 Alabama AL
     2
          Autauga County AL
     3
          Baldwin County AL
          Barbour County AL
     [5 rows x 35 columns]
[]: # Drop rows when tuitionfee out is null
     University.dropna(subset = ['tuitionfee_out'], inplace = True)
     # Calculate if a university is affordable based on the national median income_{\sqcup}
     usa_row = Unemployment[Unemployment['area_name'] == 'United States']
     national_median_income = usa_row['median_household_income_2021'].values[0]
     # interest rate for direct subsedized and unsubsidized federal loans
     annual_interest_rate = 0.055 # 5.5% annual interest rate
     monthly_interest_rate = annual_interest_rate / 12
     # federal loans are on a 10 year term
     loan_term_years = 10
     # Total number of payments
     total_payments = loan_term_years * 12
     # Calculate the monthly payment
     monthly_payment = University['tuitionfee_out'] * (monthly_interest_rate * (1 + u)
      →monthly_interest_rate)**total_payments) / ((1 +
      monthly_interest_rate)**total_payments - 1)
     # Federal government recommends student loan payments stay under 10% of total
      ⇔monthly income
     # Assumes two people in household with student loan debt
```

1

53990

53218

54762

71220

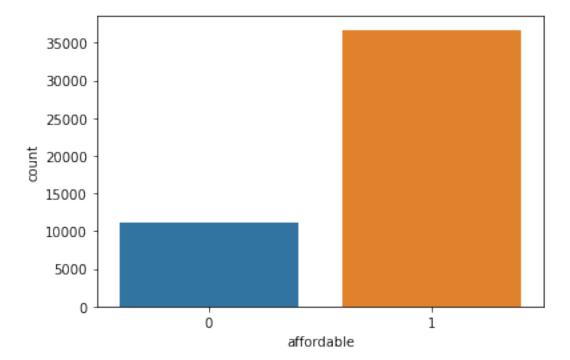
66888

```
University['affordable'] = monthly payment <= national median income / 12 * 0.05
     University['affordable'] = University['affordable'].astype(int)
[]: University.head()
[]:
        unitid
                                              instnm
                                                             city stabbr
                                                                                 zip
                                                                               35762
        100654
                           Alabama A & M University
                                                          Normal
                                                                      AL
                University of Alabama at Birmingham
     1 100663
                                                      Birmingham
                                                                      AL
                                                                          35294-0110
     2 100706
                University of Alabama in Huntsville
                                                      Huntsville
                                                                      AL
                                                                               35899
     3 100751
                          The University of Alabama
                                                      Tuscaloosa
                                                                      AL
                                                                          35487-0100
     4 100858
                                   Auburn University
                                                          Auburn
                                                                      AL
                                                                               36849
                                                  region \
     O Southeast (AL AR FL GA KY LA MS NC SC TN VA WV)
     1 Southeast (AL AR FL GA KY LA MS NC SC TN VA WV)
     2 Southeast (AL AR FL GA KY LA MS NC SC TN VA WV)
     3 Southeast (AL AR FL GA KY LA MS NC SC TN VA WV)
     4 Southeast (AL AR FL GA KY LA MS NC SC TN VA WV)
                                          preddeg
                                                   locale
                                                           latitude
                                                                      longitude
     O Predominantly bachelor's-degree granting
                                                      NaN
                                                                 NaN
                                                                            NaN
     1 Predominantly bachelor's-degree granting
                                                      NaN
                                                                 NaN
                                                                            NaN
     2 Predominantly bachelor's-degree granting
                                                      NaN
                                                                 NaN
                                                                            NaN
     3 Predominantly bachelor's-degree granting
                                                      NaN
                                                                 NaN
                                                                            NaN
     4 Predominantly bachelor's-degree granting
                                                      NaN
                                                                 NaN
                                                                            NaN
       ugds_black ugds_hisp ugds_asian
                                         ugds_aian
                                                    ugds_nhpi
                                                                ugds_2mor
                                                                           ugds_nra
     0
              NaN
                        NaN
                                    NaN
                                               NaN
                                                          NaN
                                                                      NaN
                                                                             0.0402
     1
              NaN
                        NaN
                                    NaN
                                               NaN
                                                          NaN
                                                                      NaN
                                                                             0.0330
     2
              NaN
                                    NaN
                                               NaN
                                                                             0.0396
                        NaN
                                                          NaN
                                                                      NaN
     3
              NaN
                        NaN
                                    NaN
                                                                      NaN
                                                                             0.0159
                                               NaN
                                                          NaN
     4
              NaN
                        NaN
                                    NaN
                                               NaN
                                                          NaN
                                                                      NaN
                                                                             0.0084
       ugds_unkn
                  year
                        affordable
          0.0017
                  2001
     1
          0.0255
                  2001
                                  1
     2
          0.0000
                  2001
                                  1
     3
          0.0000
                  2001
                                  1
     4
          0.0016
                  2001
                                  1
     [5 rows x 34 columns]
[]: # Show number of affordable/non-affordable universities
     value_counts = University['affordable'].value_counts()
     value_counts
```

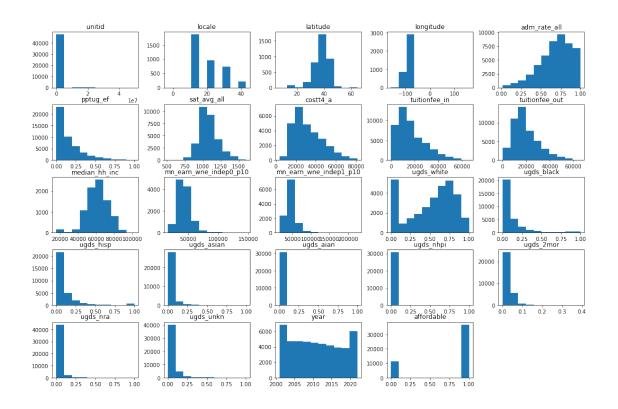
```
[]: 1 36711
0 11144
```

Name: affordable, dtype: int64

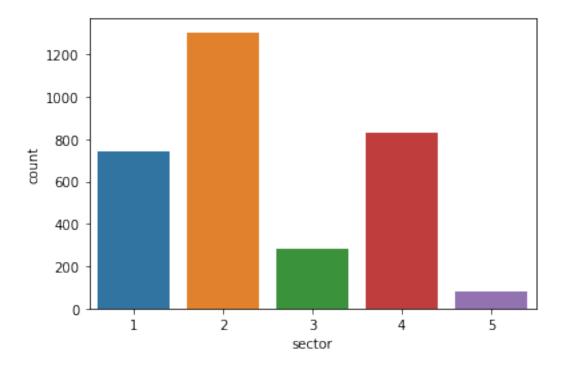
```
[]: sns.countplot(data=University, x='affordable') plt.show()
```



```
[]: University.hist(grid=False, figsize=(18,12)) plt.show()
```



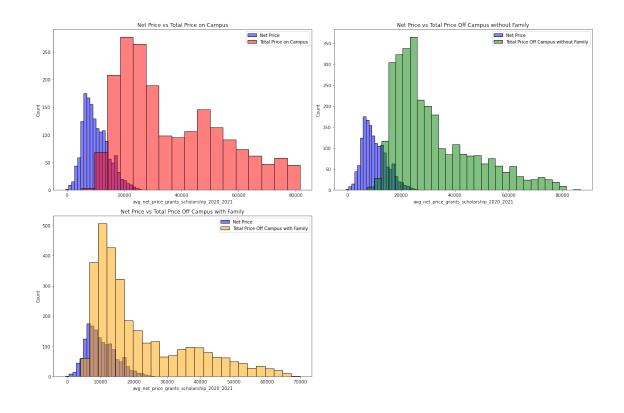
[]: sns.countplot(data=NetPrice, x='sector') plt.show()



```
[]: fig, axes = plt.subplots(2, 2, figsize=(18, 12))
    # Plot histograms in each subplot
    sns.histplot(data=NetPrice, x='avg_net_price_grants_scholarship_2020_2021',__

color='blue', alpha=0.5, label='Net Price', ax=axes[0, 0])

    sns.histplot(data=StickerPrice, x='total_price_in_state_on_campus_2020_2021',__
     ⇔color='red', alpha=0.5, label='Total Price on Campus', ax=axes[0, 0])
    axes[0, 0].legend()
    axes[0, 0].set_title('Net Price vs Total Price on Campus')
    sns.histplot(data=NetPrice, x='avg_net_price_grants_scholarship_2020_2021',_
     ⇔color='blue', alpha=0.5, label='Net Price', ax=axes[0, 1])
    sns.histplot(data=StickerPrice,_
     ⇔5, label='Total Price Off Campus without Family', ax=axes[0, 1])
    axes[0, 1].legend()
    axes[0, 1].set_title('Net Price vs Total Price Off Campus without Family')
    sns.histplot(data=NetPrice, x='avg_net_price_grants_scholarship_2020_2021',_
     ⇔color='blue', alpha=0.5, label='Net Price', ax=axes[1, 0])
    sns.histplot(data=StickerPrice,
     ⇔5, label='Total Price Off Campus with Family', ax=axes[1, 0])
    axes[1, 0].legend()
    axes[1, 0].set title('Net Price vs Total Price Off Campus with Family')
    axes[1, 1].axis('off')
    plt.tight_layout()
    plt.show()
```



[]: Unemployment.head()

[]:		fips_code	state	a	rea_name	rural	L_urban_conti	nuum_	_code_20	13	\
	0	0	US	Unite	d States				N	aN	
	1	1000	AL		Alabama				N	aN	
	2	1001	AL	Autauga C	ounty AL				2	.0	
	3	1003	AL	Baldwin C	ounty AL				3	.0	
	4	1005	AL	Barbour C	ounty AL				6	.0	
		urban_infl	luence_	_code_2013	metro_20	13 ci	ivilian_labor	forc	e_2000	\	
	0			NaN	N	aN		14260	1576.0		
	1			NaN	N	aN		214	17173.0		
	2			2.0	1	.0		2	21861.0		
	3			2.0	1	.0		6	89979.0		
	4			6.0	0	.0		1	1449.0		
		employed_2	2000 ι	nemployed_	2000 une:	mployn	ment_rate_2000)	\		
	0	1369048	53.0	56967	23.0		4.0)			
	1	204773	31.0	994	42.0		4.6	3 			
	2	2097	71.0	8	90.0		4.3	l			
	3	6737	70.0	26	09.0		3.7	7			
	4	1083	12.0	6	37.0		5.6	3 			

```
civilian_labor_force_2021
                               employed_2021 unemployed_2021 \
0
                 162229903.0
                                 153544980.0
                                                     8684923.0
1
                    2259349.0
                                   2183330.0
                                                        76019.0
2
                      26545.0
                                      25809.0
                                                          736.0
3
                      99953.0
                                      97034.0
                                                         2919.0
                       8280.0
                                       7821.0
                                                          459.0
   unemployment_rate_2021 civilian_labor_force_2022
                                                         employed_2022 \
                                           164781642.0
0
                                                           158766998.0
                       5.4
1
                       3.4
                                             2286028.0
                                                             2226670.0
2
                       2.8
                                               26789.0
                                                               26181.0
3
                       2.9
                                              102849.0
                                                              100432.0
                       5.5
                                                8241.0
                                                                7906.0
                    unemployment_rate_2022
                                             median_household_income_2021 \
   unemployed_2022
0
         6014644.0
                                                                    69717.0
                                         3.7
                                         2.6
1
           59358.0
                                                                    53990.0
2
             608.0
                                         2.3
                                                                    66444.0
3
            2417.0
                                         2.4
                                                                    65658.0
             335.0
                                         4.1
                                                                    38649.0
   med_hh_income_percent_of_state_total_2021
0
                                           NaN
                                         100.0
1
2
                                         123.1
3
                                         121.6
                                          71.6
```

[5 rows x 100 columns]

2.4 2.1 Join Datasets

```
uni.MN_EARN_WNE_INDEP1_P10,
        uni.PREDDEG,
        uni.CONTROL,
        pov.Stabr AS state_abbr,
        pov.Area_name_with_state,
        pov.POVALL_2021,
        pov.CI90LBALL 2021,
        pov.CI90UBALL 2021,
        pov.MEDHHINC 2021,
        pov.CI90LBINC 2021,
        pov.CI90UBINC 2021.
        unemp. Unemployment rate 2000,
        unemp. Unemployment rate 2001,
        unemp.Unemployment_rate_2002,
        unemp. Unemployment rate 2003,
        unemp.Unemployment_rate_2004,
        unemp.Unemployment_rate_2005,
        unemp. Unemployment rate 2006,
        unemp.Unemployment_rate_2007,
        unemp.Unemployment_rate_2008,
        unemp.Unemployment_rate_2009,
        unemp. Unemployment rate 2010,
        unemp. Unemployment rate 2011,
        unemp. Unemployment rate 2013,
        unemp. Unemployment rate 2014,
        unemp. Unemployment rate 2015,
        unemp. Unemployment rate 2016,
        unemp. Unemployment rate 2017,
        unemp.Unemployment_rate_2018,
        unemp. Unemployment rate 2019,
        unemp.Unemployment_rate_2020,
        unemp.Unemployment_rate_2021,
        unemp. Unemployment rate 2022,
        unemp.Median_Household_Income_2021,
        net.avg_net_price_grants_scholarship_2020_2021,
        net.avg_net_price_grants_scholarship_2019 2020,
        net.avg net price grants scholarship 2018 2019,
        net.avg_net_price_grants_scholarship_2017_2018,
        net.avg_net_price_grants_scholarship_2016_2017,
        net.avg net price grants scholarship 2015 2016,
        net.avg net price grants scholarship 2014 2015,
        net.avg_net_price_grants_scholarship_2013_2014,
        net.avg_net_price_grants_scholarship_2012_2013,
        net.avg_net_price_grants_scholarship_2011_2012,
        unemp.Unemployment_rate_2012,
        uni.affordable
FROM University uni
```

```
JOIN Unemployment unemp ON pov.FIPS_Code = unemp.FIPS_Code
     JOIN NetPrice net ON uni.INSTNM = net.institution_name
     WHERE uni.year = 2022
     0.00
     # Execute the SQL query on the Pandas DataFrames
     df = sqldf(combine_tables_query, globals())
     # Display the result
     df.head()
[]:
       unitid
                                  instnm stabbr
                                                   city \
     0 100654 Alabama A & M University
                                             AL Normal
     1 100654 Alabama A & M University
                                             AL
                                                 Normal
     2 100654 Alabama A & M University
                                             AL Normal
     3 100654 Alabama A & M University
                                             AL Normal
     4 100654 Alabama A & M University
                                             AL Normal
                                                 region
                                                         adm_rate_all sat_avg_all \
     O Southeast (AL AR FL GA KY LA MS NC SC TN VA WV)
                                                              0.716006
                                                                              954.0
     1 Southeast (AL AR FL GA KY LA MS NC SC TN VA WV)
                                                             0.716006
                                                                              954.0
     2 Southeast (AL AR FL GA KY LA MS NC SC TN VA WV)
                                                             0.716006
                                                                              954.0
     3 Southeast (AL AR FL GA KY LA MS NC SC TN VA WV)
                                                             0.716006
                                                                              954.0
     4 Southeast (AL AR FL GA KY LA MS NC SC TN VA WV)
                                                             0.716006
                                                                              954.0
       costt4_a tuitionfee_in tuitionfee_out
     0
        21924.0
                        10024.0
                                        18634.0 ...
         21924.0
                        10024.0
     1
                                        18634.0 ...
         21924.0
     2
                        10024.0
                                        18634.0 ...
     3
         21924.0
                        10024.0
                                        18634.0 ...
         21924.0
                        10024.0
                                        18634.0 ...
       avg_net_price_grants_scholarship_2018_2019
     0
                                           14604.0
     1
                                           14604.0
     2
                                           14604.0
     3
                                           14604.0
     4
                                           14604.0
       avg_net_price_grants_scholarship_2017_2018 \
     0
                                          13956.0
                                          13956.0
     1
     2
                                          13956.0
     3
                                          13956.0
     4
                                          13956.0
```

JOIN Poverty pov ON uni.STABBR = pov.Stabr

```
avg_net_price_grants_scholarship_2016_2017 \
0
                                        15812.0
                                         15812.0
1
2
                                         15812.0
3
                                        15812.0
4
                                         15812.0
  avg\_net\_price\_grants\_scholarship\_2015\_2016 \quad \setminus
0
                                         15547.0
1
                                         15547.0
2
                                         15547.0
3
                                        15547.0
4
                                         15547.0
  avg_net_price_grants_scholarship_2014_2015
                                        13203.0
0
                                        13203.0
1
2
                                         13203.0
3
                                        13203.0
4
                                         13203.0
  avg_net_price_grants_scholarship_2013_2014
0
                                         14746.0
                                         14746.0
1
2
                                         14746.0
3
                                         14746.0
4
                                         14746.0
  avg_net_price_grants_scholarship_2012_2013
0
                                        12887.0
1
                                        12887.0
2
                                         12887.0
3
                                         12887.0
4
                                         12887.0
  avg_net_price_grants_scholarship_2011_2012 unemployment_rate_2012 affordable
                                        11108.0
                                                                      8.2
0
                                                                                    1
                                                                                    1
1
                                        11108.0
                                                                      7.1
2
                                        11108.0
                                                                      7.7
                                                                                    1
3
                                        11108.0
                                                                     11.8
                                                                                    1
                                        11108.0
                                                                      8.8
```

[5 rows x 56 columns]

3 3. Exploratory Data Analysis

```
[]: # get number of rows and columns
    print('Number of Rows in University - ', University.shape[0])
    print('Number of Columns in University - ', University.shape[1], '\n')
    print('Number of Rows in NetPrice - ', NetPrice.shape[0])
    print('Number of Columns in NetPrice - ', NetPrice.shape[1], '\n')
    print('Number of Rows in Poverty - ', Poverty.shape[0])
    print('Number of Columns in Poverty - ', Poverty.shape[1], '\n')
    print('Number of Rows in Unemployment - ', Unemployment.shape[0])
    print('Number of Columns in Unemployment - ', Unemployment.shape[1], '\n')
    Number of Rows in University - 47855
    Number of Columns in University - 34
    Number of Rows in NetPrice - 3240
    Number of Columns in NetPrice - 63
    Number of Rows in Poverty - 3195
    Number of Columns in Poverty - 35
    Number of Rows in Unemployment - 3277
    Number of Columns in Unemployment - 100
```

3.1 3.1 Data Types and NULLS

Number of Rows - 137872 Number of Columns - 56

[]:	Field	Data Type	Nulls
0	unitid	int64	0
1	instnm	object	0
2	stabbr	object	0
3	city	object	0
4	region	object	0
5	adm_rate_all	float64	0
6	sat_avg_all	float64	48619
7	costt4_a	float64	4001
8	tuitionfee_in	float64	0
9	tuitionfee_out	float64	0
10	mn_earn_wne_indep1_p10	float64	36033
11	preddeg	object	0
12	control	object	0
13	state_abbr	object	0
14	area_name_with_state	object	0
15	povall_2021	object	0
16	ci90lball_2021	object	0
17	ci90uball_2021	object	0
18 19	medhhinc_2021 ci90lbinc_2021	object object	0
20	ci90ubinc_2021	object	0
21	unemployment_rate_2000	float64	28
22	unemployment_rate_2001	float64	28
23	unemployment_rate_2002	float64	28
24	unemployment_rate_2003	float64	28
25	unemployment_rate_2004	float64	28
26	unemployment_rate_2005	float64	196
27	unemployment_rate_2006	float64	196
28	unemployment_rate_2007	float64	28
29	unemployment_rate_2008	float64	28
30	unemployment_rate_2009	float64	28
31	unemployment_rate_2010	float64	8
32	unemployment_rate_2011	float64	8
33	unemployment_rate_2013	float64	8
34	unemployment_rate_2014	float64	8
35	unemployment_rate_2015	float64	8
36	unemployment_rate_2016	float64	8
37	unemployment_rate_2017	float64	8
38	unemployment_rate_2018	float64 float64	8
39 40	unemployment_rate_2019 unemployment_rate_2020	float64	8
41	unemployment_rate_2020 unemployment_rate_2021	float64	0
41	unemployment_rate_2021 unemployment_rate_2022	float64	0
43	median_household_income_2021	float64	0
44	avg_net_price_grants_scholarship_2020_2021	float64	92267
45	avg_net_price_grants_scholarship_2019_2020	float64	92267
	0_ 1_I 1 1 0 1 0 1 0 1 0 0 0 0 0 0 0 0 0 0 0	·	

```
float64 92562
46
   avg_net_price_grants_scholarship_2018_2019
47
   avg_net_price_grants_scholarship_2017_2018
                                                float64 92562
   avg_net_price_grants_scholarship_2016_2017
                                                float64
                                                         92562
48
   avg_net_price_grants_scholarship_2015_2016
49
                                                float64
                                                         92919
50
   avg_net_price_grants_scholarship_2014_2015
                                                float64 92956
   avg_net_price_grants_scholarship_2013_2014
                                                float64 93610
51
   avg_net_price_grants_scholarship_2012_2013
52
                                                float64 93966
    avg_net_price_grants_scholarship_2011_2012
                                                float64 94446
53
54
                       unemployment_rate_2012
                                                float64
                                                             8
55
                                   affordable
                                                  int64
                                                             0
```

3.2 3.2 Summary Statistics

```
[]: #summary statistics
stats = pd.DataFrame(df.describe()).T
stats
```

[]:	count	mean	\
unitid	137872.0	215348.687435	
adm_rate_all	137872.0	0.733948	
sat_avg_all	89253.0	1151.858660	
costt4_a	133871.0	36742.799083	
tuitionfee_in	137872.0	23755.840555	
tuitionfee_out	137872.0	27619.961624	
<pre>mn_earn_wne_indep1_p10</pre>	101839.0	50801.502371	
unemployment_rate_2000	137844.0	4.396701	
unemployment_rate_2001	137844.0	5.086958	
unemployment_rate_2002	137844.0	5.928520	
unemployment_rate_2003	137844.0	6.181507	
unemployment_rate_2004	137844.0	5.852187	
unemployment_rate_2005	137676.0	5.534203	
unemployment_rate_2006	137676.0	5.077598	
unemployment_rate_2007	137844.0	5.042835	
unemployment_rate_2008	137844.0	6.101547	
unemployment_rate_2009	137844.0	9.488022	
unemployment_rate_2010	137864.0	9.768224	
unemployment_rate_2011	137864.0	9.078855	
unemployment_rate_2013	137864.0	7.650666	
unemployment_rate_2014	137864.0	6.384567	
unemployment_rate_2015	137864.0	5.635530	
unemployment_rate_2016	137864.0	5.361369	
unemployment_rate_2017	137864.0	4.746569	
unemployment_rate_2018	137864.0	4.199532	
unemployment_rate_2019	137864.0	3.996497	
unemployment_rate_2020	137872.0	7.208137	
unemployment_rate_2021	137872.0	4.964397	
unemployment_rate_2022	137872.0	3.758887	

```
median_household_income_2021
                                              137872.0
                                                         60422.055421
                                               45605.0
                                                         13594.144392
avg_net_price_grants_scholarship_2020_2021
avg_net_price_grants_scholarship_2019_2020
                                               45605.0
                                                         13609.008990
avg_net_price_grants_scholarship_2018_2019
                                               45310.0
                                                         13320.887641
avg_net_price_grants_scholarship_2017_2018
                                               45310.0
                                                         13175.989031
avg_net_price_grants_scholarship_2016_2017
                                               45310.0
                                                         13104.512094
avg_net_price_grants_scholarship_2015_2016
                                               44953.0
                                                         12762.443441
avg_net_price_grants_scholarship_2014_2015
                                               44916.0
                                                         12554.149323
avg net price grants scholarship 2013 2014
                                               44262.0
                                                         12100.449415
avg_net_price_grants_scholarship_2012_2013
                                               43906.0
                                                         12024.602993
avg net price grants scholarship 2011 2012
                                               43426.0
                                                         11692.709943
unemployment_rate_2012
                                              137864.0
                                                             8.201889
affordable
                                              137872.0
                                                             0.528983
                                                       std
                                                                       min
                                                            100654.000000
unitid
                                              93979.963627
                                                                 0.039152
adm_rate_all
                                                  0.206004
                                                               776.000000
sat_avg_all
                                                129.362088
costt4_a
                                              16320.646884
                                                              8118.000000
tuitionfee_in
                                              15257.251778
                                                               480.000000
tuitionfee_out
                                              13156.366922
                                                               480.000000
mn_earn_wne_indep1_p10
                                              15799.219146
                                                             23400.000000
unemployment_rate_2000
                                                  1.598401
                                                                  1.300000
unemployment rate 2001
                                                                  1.600000
                                                  1.707137
unemployment rate 2002
                                                  1.811223
                                                                  1.600000
unemployment rate 2003
                                                  1.866708
                                                                  1.900000
unemployment_rate_2004
                                                  1.734494
                                                                  1.600000
unemployment rate 2005
                                                  1.702856
                                                                 2.000000
unemployment_rate_2006
                                                  1.613475
                                                                  1.600000
unemployment rate 2007
                                                                  1.400000
                                                  1.621889
unemployment_rate_2008
                                                  1.999010
                                                                  1.300000
unemployment_rate_2009
                                                                 2.000000
                                                  3.031887
unemployment rate 2010
                                                                 2.000000
                                                  2.918881
unemployment_rate_2011
                                                  2.776444
                                                                  1.400000
unemployment_rate_2013
                                                  2.388310
                                                                  1,200000
unemployment_rate_2014
                                                  2.050799
                                                                  1.200000
unemployment rate 2015
                                                  1.813046
                                                                  1.800000
unemployment rate 2016
                                                  1.746241
                                                                  1.600000
unemployment rate 2017
                                                  1.511904
                                                                  1.500000
unemployment rate 2018
                                                                  1.200000
                                                  1.348143
unemployment rate 2019
                                                  1.330921
                                                                 0.800000
unemployment_rate_2020
                                                  2.207568
                                                                  1.600000
unemployment rate 2021
                                                                 0.900000
                                                  1.689318
unemployment_rate_2022
                                                  1.188747
                                                                 0.600000
                                                             25653.000000
median_household_income_2021
                                              15836.881547
avg_net_price_grants_scholarship_2020_2021
                                               4090.581038
                                                              2695.000000
avg_net_price_grants_scholarship_2019_2020
                                               4111.857018
                                                              2958.000000
```

```
avg_net_price_grants_scholarship_2018_2019
                                               4249.506289
                                                              2465.000000
avg net price grants scholarship 2017 2018
                                               4177.775661
                                                              2420.000000
avg_net_price_grants_scholarship_2016_2017
                                               4023.185233
                                                              2304.000000
avg_net_price_grants_scholarship_2015_2016
                                               3675.339126
                                                              3156.000000
avg_net_price_grants_scholarship_2014_2015
                                               3786.272570
                                                              2345.000000
avg_net_price_grants_scholarship_2013_2014
                                               3494.014386
                                                              1640.000000
avg_net_price_grants_scholarship_2012_2013
                                               3325.762179
                                                              1993.000000
avg_net_price_grants_scholarship_2011_2012
                                               3422.225105
                                                              1323.000000
unemployment rate 2012
                                                  2.614596
                                                                  1.100000
affordable
                                                  0.499161
                                                                  0.00000
                                                        25%
                                                                        50%
unitid
                                              155812.000000
                                                             198899.000000
adm_rate_all
                                                   0.626667
                                                                   0.776997
                                                1064.000000
                                                                1127.000000
sat_avg_all
costt4_a
                                               23078.000000
                                                              33560.000000
                                                              19469.000000
tuitionfee_in
                                               10044.000000
tuitionfee_out
                                               17488.000000
                                                              25570.000000
                                                              48300.000000
mn_earn_wne_indep1_p10
                                               41200.000000
unemployment_rate_2000
                                                   3,400000
                                                                   4.100000
unemployment_rate_2001
                                                   4.000000
                                                                   4.800000
unemployment rate 2002
                                                   4.800000
                                                                   5.700000
unemployment_rate_2003
                                                   4.900000
                                                                   5.900000
unemployment rate 2004
                                                   4.700000
                                                                   5.600000
unemployment rate 2005
                                                   4.400000
                                                                   5.300000
unemployment rate 2006
                                                   4.100000
                                                                   4.900000
unemployment_rate_2007
                                                   4.000000
                                                                   4.800000
unemployment rate 2008
                                                   4.800000
                                                                   5.900000
unemployment_rate_2009
                                                   7.400000
                                                                   9.100000
unemployment rate 2010
                                                   7.800000
                                                                   9.500000
unemployment_rate_2011
                                                   7.300000
                                                                   8.800000
unemployment_rate_2013
                                                   6.100000
                                                                   7.500000
unemployment rate 2014
                                                   5.000000
                                                                   6.200000
unemployment_rate_2015
                                                   4.400000
                                                                   5.400000
unemployment_rate_2016
                                                   4.300000
                                                                   5.100000
unemployment_rate_2017
                                                   3.700000
                                                                   4.500000
unemployment rate 2018
                                                   3.300000
                                                                   4.000000
unemployment rate 2019
                                                   3.100000
                                                                   3.800000
unemployment rate 2020
                                                   5.700000
                                                                   7.100000
unemployment rate 2021
                                                   3.900000
                                                                   4.700000
unemployment rate 2022
                                                   3.000000
                                                                   3.600000
median_household_income_2021
                                               50109.000000
                                                              57584.000000
avg_net_price_grants_scholarship_2020_2021
                                               11076.000000
                                                               13639.000000
avg_net_price_grants_scholarship_2019_2020
                                               11175.000000
                                                              13481.000000
avg_net_price_grants_scholarship_2018_2019
                                               10455.000000
                                                              13212.000000
avg_net_price_grants_scholarship_2017_2018
                                               10491.000000
                                                              12913.000000
                                               10895.000000
avg_net_price_grants_scholarship_2016_2017
                                                               13017.000000
```

<pre>avg_net_price_grants_scholarship_2015_2016</pre>	10330.000000	12647.000000
<pre>avg_net_price_grants_scholarship_2014_2015</pre>	10166.000000	12550.000000
<pre>avg_net_price_grants_scholarship_2013_2014</pre>	10003.000000	12046.000000
<pre>avg_net_price_grants_scholarship_2012_2013</pre>	9971.000000	11873.000000
<pre>avg_net_price_grants_scholarship_2011_2012</pre>	9632.000000	11792.000000
unemployment_rate_2012	6.500000	8.000000
affordable	0.000000	1.000000
	75%	max
unitid	227331.000000	497268.0
adm_rate_all	0.890354	1.0
sat_avg_all	1228.000000	1537.0
costt4_a	47588.000000	81531.0
tuitionfee_in	34481.000000	66064.0
tuitionfee_out	35500.000000	66064.0
mn_earn_wne_indep1_p10	55700.000000	224600.0
unemployment_rate_2000	5.100000	17.3
unemployment_rate_2001	5.800000	17.6
unemployment_rate_2002	6.800000	19.6
unemployment_rate_2003	7.100000	17.8
unemployment_rate_2004	6.600000	20.2
unemployment_rate_2005	6.300000	21.0
unemployment_rate_2006	5.800000	20.7
unemployment_rate_2007	5.700000	20.2
unemployment_rate_2008	7.100000	22.6
unemployment_rate_2009	11.200000	28.3
unemployment_rate_2010	11.500000	29.4
unemployment_rate_2011	10.500000	29.3
unemployment_rate_2013	9.000000	25.6
unemployment_rate_2014	7.500000	24.3
unemployment_rate_2015	6.500000	24.6
unemployment_rate_2016	6.100000	24.2
unemployment_rate_2017	5.500000	19.7
unemployment_rate_2018	4.900000	18.8
unemployment_rate_2019	4.600000	20.7
unemployment_rate_2020	8.400000	22.6
unemployment_rate_2021	5.800000	19.5
unemployment_rate_2022	4.300000	14.7
median_household_income_2021	66601.000000	153716.0
<pre>avg_net_price_grants_scholarship_2020_2021</pre>	16433.000000	26179.0
<pre>avg_net_price_grants_scholarship_2019_2020</pre>	16166.000000	27675.0
<pre>avg_net_price_grants_scholarship_2018_2019</pre>	15886.000000	40090.0
<pre>avg_net_price_grants_scholarship_2017_2018</pre>	15636.000000	44661.0
avg_net_price_grants_scholarship_2016_2017	15664.000000	40927.0
avg_net_price_grants_scholarship_2015_2016	15235.000000	25097.0
avg_net_price_grants_scholarship_2014_2015	14625.000000	41647.0
<pre>avg_net_price_grants_scholarship_2013_2014</pre>	14418.000000	25138.0

```
      avg_net_price_grants_scholarship_2012_2013
      14299.000000
      24247.0

      avg_net_price_grants_scholarship_2011_2012
      13890.000000
      24674.0

      unemployment_rate_2012
      9.600000
      27.7

      affordable
      1.000000
      1.0
```

Outliers with Z-score (Threshold 3)

```
[]: # In-state tuition data
     tuitionfee_in_data = df['tuitionfee_in']
     # Z-scores
     z_scores_in = (tuitionfee_in_data - np.mean(tuitionfee_in_data)) / np.
      ⇔std(tuitionfee_in_data)
     threshold = 3
     # Find indices of outliers
     outliers_in = np.where(np.abs(z_scores_in) > threshold)[0]
     # Detail information about outliers
     outlier_details = df.iloc[outliers_in]
     # Print information about outliers
     print("Outliers in tuitionfee_in:")
     print(outlier_details)
     # Summary statistics
     print("\nSummary Statistics:")
     print("Mean Tuition Fee:", np.mean(tuitionfee_in_data))
     print("Median Tuition Fee:", np.median(tuitionfee_in_data))
     print("Standard Deviation:", np.std(tuitionfee_in_data))
     print("Number of Outliers:", len(outliers_in))
```

Outliers in tuitionfee_in:

Empty DataFrame

```
Columns: [unitid, instnm, stabbr, city, region, adm_rate_all, sat_avg_all, costt4_a, tuitionfee_in, tuitionfee_out, mn_earn_wne_indep1_p10, preddeg, control, state_abbr, area_name_with_state, povall_2021, ci90lball_2021, ci90uball_2021, medhhinc_2021, ci90lbinc_2021, ci90ubinc_2021, unemployment_rate_2000, unemployment_rate_2001, unemployment_rate_2002, unemployment_rate_2003, unemployment_rate_2004, unemployment_rate_2005, unemployment_rate_2006, unemployment_rate_2007, unemployment_rate_2008, unemployment_rate_2009, unemployment_rate_2010, unemployment_rate_2011, unemployment_rate_2013, unemployment_rate_2014, unemployment_rate_2015, unemployment_rate_2016, unemployment_rate_2017, unemployment_rate_2018, unemployment_rate_2019, unemployment_rate_2020, unemployment_rate_2021, unemployment_rate_2022, median_household_income_2021, avg_net_price_grants_scholarship_2020_2021, avg_net_price_grants_scholarship_2020_2020,
```

```
avg_net_price_grants_scholarship_2018_2019,
avg_net_price_grants_scholarship_2017_2018,
avg_net_price_grants_scholarship_2016_2017,
avg_net_price_grants_scholarship_2015_2016,
avg_net_price_grants_scholarship_2014_2015,
avg_net_price_grants_scholarship_2013_2014,
avg_net_price_grants_scholarship_2012_2013,
avg_net_price_grants_scholarship_2011_2012, unemployment_rate_2012, affordable]
Index: []
```

[0 rows x 56 columns]

Summary Statistics:

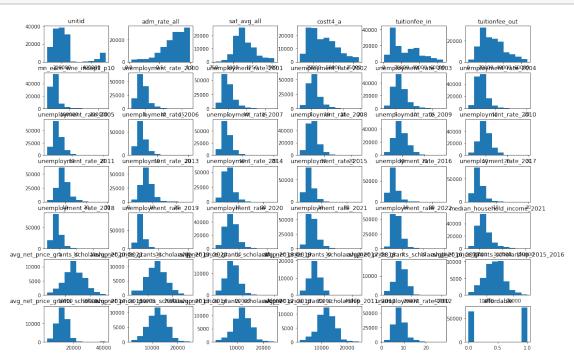
Mean Tuition Fee: 23755.840555007544

Median Tuition Fee: 19469.0

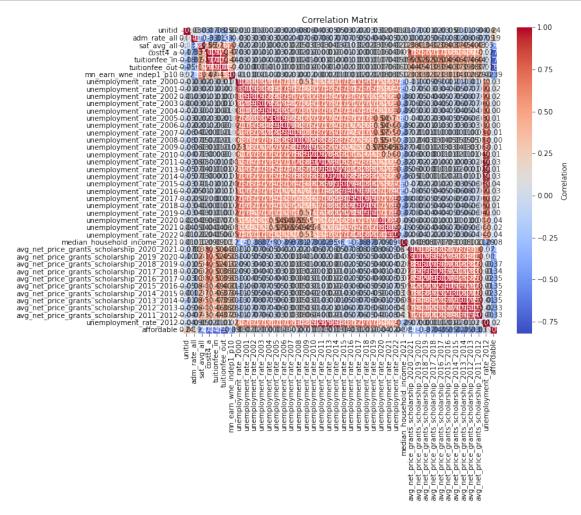
Standard Deviation: 15257.196447110557

Number of Outliers: 0

[]: df.hist(grid=False, figsize=(18,12)) plt.show()



3.3 Multicollinearity Test



```
[]: # This is not representative of what we will drop yet. Of course in-state and out-of-state tutition of correlated, but will not be dropped.

# Included it in case we find something interesting on joined table
```

```
[]: # Calculate the correlation of affordability with all other attributes
     correlation_with_target = df.corrwith(df['affordable'])
     # Define correlation thresholds
     upper_threshold = 0.4
     lower_threshold = -0.4
     # Get attributes above and below thresholds
     above threshold = correlation with target[correlation with target.abs() > 1
      →upper_threshold]
     below_threshold = correlation_with_target[correlation_with_target.abs() <__
      →lower_threshold]
     # Show most correlated attributes
     print(above threshold)
     print(below_threshold)
                                                  -0.471855
    sat_avg_all
    costt4 a
                                                  -0.743210
    tuitionfee_in
                                                  -0.751890
    tuitionfee out
                                                  -0.821012
    avg_net_price_grants_scholarship_2020_2021
                                                  -0.408468
    avg_net_price_grants_scholarship_2019_2020
                                                 -0.407662
    affordable
                                                   1.000000
    dtype: float64
    Series([], dtype: float64)
[]: cor_matrix = df.corr().abs()
     upper_tri = cor_matrix.where(np.triu(np.ones(cor_matrix.shape),
                                          k=1).astype(bool))
     to_drop = [column for column in upper_tri.columns if
                any(upper tri[column] > 0.80)]
     print('Columns with chance of multicollinearity: %s'%to_drop)
    Columns with chance of multicollinearity: ['tuitionfee_in', 'tuitionfee_out',
    'unemployment_rate_2001', 'unemployment_rate_2002', 'unemployment_rate_2003',
    'unemployment_rate_2004', 'unemployment_rate_2005', 'unemployment_rate_2006',
    'unemployment rate 2007', 'unemployment rate 2008', 'unemployment rate 2009',
    'unemployment_rate_2010', 'unemployment_rate_2011', 'unemployment_rate_2013',
    'unemployment_rate_2014', 'unemployment_rate_2015', 'unemployment_rate_2016',
    'unemployment_rate_2017', 'unemployment_rate_2018', 'unemployment_rate_2019',
    'unemployment_rate_2021', 'unemployment_rate_2022',
    'avg_net_price_grants_scholarship_2019_2020',
    'avg_net_price_grants_scholarship_2018_2019',
    'avg_net_price_grants_scholarship_2017_2018',
    'avg_net_price_grants_scholarship_2016_2017',
```

```
'avg_net_price_grants_scholarship_2015_2016',
'avg_net_price_grants_scholarship_2014_2015',
'avg_net_price_grants_scholarship_2013_2014',
'avg_net_price_grants_scholarship_2012_2013',
'avg_net_price_grants_scholarship_2011_2012', 'unemployment_rate_2012',
'affordable']
```

4 4. Data Preparation

4.0.1 Handle Duplicate, Null, and Multicollinearity Features

```
[]: # Drop redundant rows
     df = df.drop_duplicates()
[]: # Get null values for each column in the university dataset
     null_cols = df.isnull().sum()
     total_rows = df['instnm'].count()
     percent_col_null = null_cols/total_rows * 100
     percent_col_null.sort_values(ascending=False)
[]: avg_net_price_grants_scholarship_2011_2012
                                                    68.502669
     avg_net_price_grants_scholarship_2012 2013
                                                    68.154520
     avg_net_price_grants_scholarship_2013_2014
                                                    67.896310
     avg_net_price_grants_scholarship_2014_2015
                                                    67.421957
     avg_net_price_grants_scholarship_2015_2016
                                                    67.395120
     avg_net_price_grants_scholarship_2016_2017
                                                    67.136184
     avg_net_price_grants_scholarship_2017_2018
                                                    67.136184
     avg_net_price_grants_scholarship_2018_2019
                                                    67.136184
     avg_net_price_grants_scholarship_2019_2020
                                                    66.922218
     avg_net_price_grants_scholarship_2020_2021
                                                    66.922218
                                                    35.263868
     sat_avg_all
    mn_earn_wne_indep1_p10
                                                    26.135111
     costt4 a
                                                     2.901967
     unemployment_rate_2005
                                                     0.142161
     unemployment rate 2006
                                                     0.142161
     unemployment_rate_2008
                                                     0.020309
     unemployment rate 2009
                                                     0.020309
     unemployment_rate_2000
                                                     0.020309
     unemployment_rate_2001
                                                     0.020309
     unemployment_rate_2002
                                                     0.020309
     unemployment_rate_2003
                                                     0.020309
     unemployment_rate_2004
                                                     0.020309
     unemployment rate 2007
                                                     0.020309
     unemployment rate 2014
                                                     0.005802
     unemployment rate 2013
                                                     0.005802
    unemployment rate 2011
                                                     0.005802
     unemployment_rate_2010
                                                     0.005802
```

```
unemployment_rate_2015
                                                     0.005802
     unemployment_rate_2016
                                                      0.005802
     unemployment_rate_2018
                                                     0.005802
     unemployment_rate_2019
                                                      0.005802
     unemployment_rate_2017
                                                     0.005802
     unemployment_rate_2012
                                                     0.005802
     unemployment_rate_2020
                                                     0.000000
     unemployment_rate_2022
                                                     0.000000
     median household income 2021
                                                     0.000000
     unemployment_rate_2021
                                                     0.000000
     unitid
                                                     0.000000
     instnm
                                                     0.000000
     ci90ubinc 2021
                                                     0.000000
     stabbr
                                                     0.000000
                                                     0.000000
     city
                                                     0.000000
     region
     adm_rate_all
                                                     0.000000
     tuitionfee_in
                                                     0.000000
     tuitionfee_out
                                                     0.000000
     preddeg
                                                      0.000000
     control
                                                     0.000000
     state abbr
                                                     0.000000
     area_name_with_state
                                                     0.000000
     povall 2021
                                                     0.000000
     ci901ball_2021
                                                     0.000000
     ci90uball 2021
                                                     0.000000
    medhhinc_2021
                                                     0.000000
     ci901binc 2021
                                                     0.000000
     affordable
                                                     0.000000
     dtype: float64
[]: # Get the size of the dataframe before dropping null values
     df.shape
[]: (137872, 56)
[]: # Drop null values
     df = df.dropna()
[]: # Get the size of the dataframe after dropping null values
     df.shape
[]: (32944, 56)
```

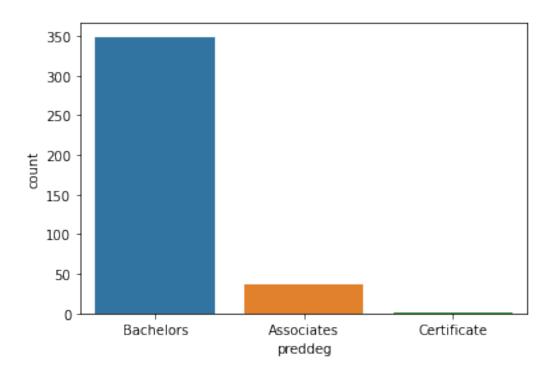
4.1 4.3 Additional Visualizations to Understand Bias

```
[]: preddeg count
0 Predominantly bachelor's-degree granting 349
1 Predominantly associate's-degree granting 36
2 Predominantly certificate-degree granting 2
```

4.1.1 4.3.1 Barplot of Primary Degree Granted at Universities

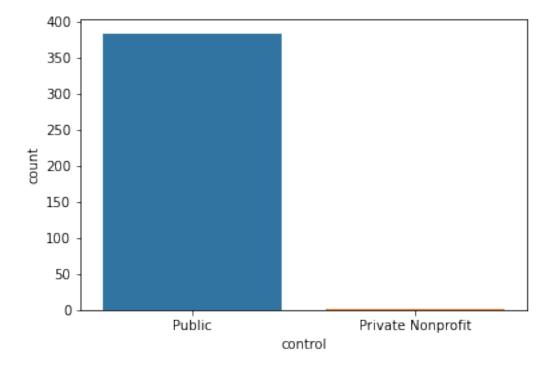
```
[]: degree_labels = ['Bachelors', 'Associates', 'Certificate']

sns.barplot(data=degree_count, x='preddeg', y='count')
plt.xticks(ticks=range(len(degree_labels)), labels=degree_labels)
plt.show()
```



```
[]: control count
0 Public 384
1 Private nonprofit 1
```

4.1.2 4.3.2 Barplot of University Funding Control



```
[]: # Define all U.S. regions. Since these are cetegories, we might consider using 

→ one-hot encoder

regions = {'Southeast (AL AR FL GA KY LA MS NC SC TN VA WV)':'Southeast',

'Great Lakes (IL IN MI OH WI)':'Great Lakes',

'Southwest (AZ NM OK TX)':'Southwest',

'Mid East (DE DC MD NJ NY PA)':'Mid East',

'Plains (IA KS MN MO NE ND SD)':'Plains',

'Far West (AK CA HI NV OR WA)':'Far East',

'Rocky Mountains (CO ID MT UT WY)':'Rocky Mountains',
```

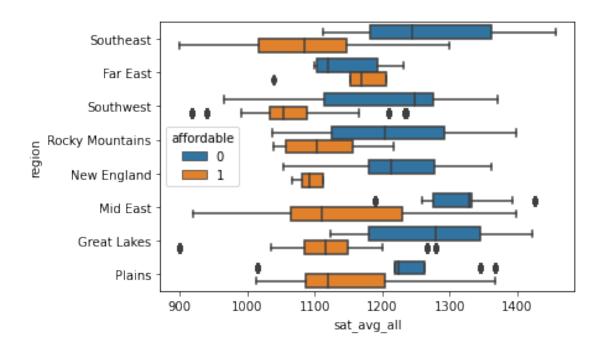
```
'New England (CT ME MA NH RI VT)':'New England',
    'U.S. Service Schools':'U.S. Service Schools'}

df['region'] = df['region'].replace(regions)
```

```
[]:
                 region count
              Southeast
                           119
            Great Lakes
     1
                            77
     2
               Mid East
                            53
              Southwest
     3
                            47
     4
                 Plains
                            39
            New England
     5
                            22
     6 Rocky Mountains
                            18
     7
               Far East
                            10
```

4.1.3 4.3.3 Boxplot of U.S. Regions, SAT Scores, and Affordability

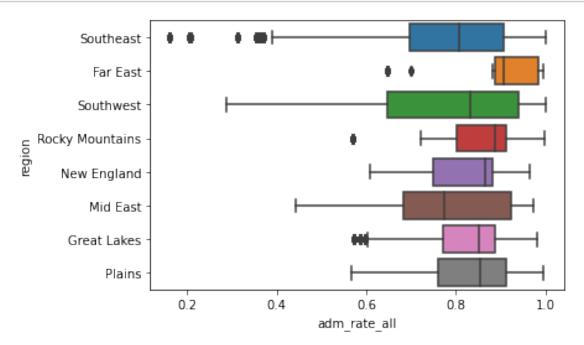
```
[]: sns.boxplot(data=df, y='region', x='sat_avg_all', hue='affordable')
plt.figure(figsize=(10, 8))
plt.show()
```



<Figure size 720x576 with 0 Axes>

4.1.4 4.3.4 Boxplot of U.S. Regions and Admissions Rate

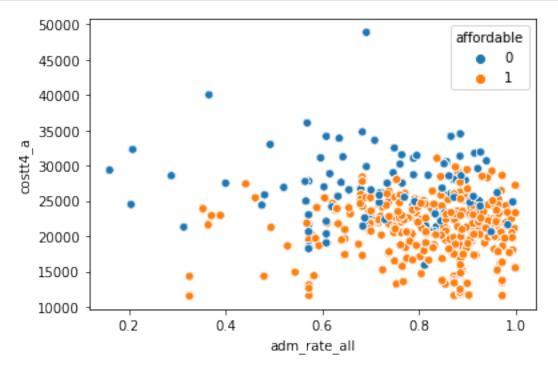
```
[]: sns.boxplot(data=df, y='region', x='adm_rate_all')
plt.figure(figsize=(10, 8))
plt.show()
```



<Figure size 720x576 with 0 Axes>

4.1.5 4.3.5 Scatterplot of Admissions Rate, Overall Cost of Tuition, and Affordability

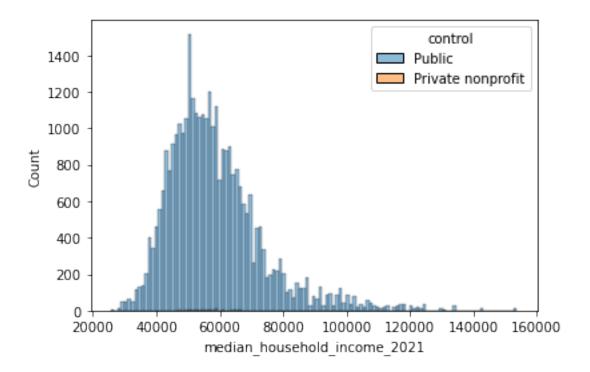
```
[]: sns.scatterplot(data=df, x='adm_rate_all', y='costt4_a', hue='affordable')
plt.figure(figsize=(10, 8))
plt.show()
```



<Figure size 720x576 with 0 Axes>

4.1.6 4.3.6 Histogram of 2021 Median Household Income and Enrollment in University Funding Control

```
[]: sns.histplot(data=df, x='median_household_income_2021', hue='control')
plt.figure(figsize=(10, 8))
plt.show()
```



<Figure size 720x576 with 0 Axes>

```
[]: import os

# Define the folder name
folder_name = 'data'

# Check if the folder exists, if not, create it
if not os.path.exists(folder_name):
    os.makedirs(folder_name)

# Save the DataFrame to a CSV file inside the folder
df.to_csv(os.path.join(folder_name, 'combined_data.csv'), index=False)
```

4.2 4.4 Encoding Categorical Labels

```
[]: from sklearn.compose import ColumnTransformer from sklearn.preprocessing import OneHotEncoder
```

```
# Encoding U.S. regions
     encoded_df = pd.get_dummies(df, columns=['region'])
     encoded_df.head()
[]:
                       sat_avg_all costt4_a tuitionfee_in tuitionfee_out \
        adm_rate_all
     0
            0.716006
                             954.0
                                      21924.0
                                                      10024.0
                                                                       18634.0
     1
            0.716006
                             954.0
                                      21924.0
                                                      10024.0
                                                                       18634.0
     2
            0.716006
                             954.0
                                      21924.0
                                                      10024.0
                                                                       18634.0
                             954.0
     3
            0.716006
                                      21924.0
                                                      10024.0
                                                                       18634.0
            0.716006
                             954.0
                                      21924.0
                                                      10024.0
                                                                       18634.0
        mn_earn_wne_indep1_p10 preddeg povall_2021 ci90lball_2021 \
     0
                        38100.0
                                        2
                                                800848
                                                                 782169
                                        2
     1
                        38100.0
                                                   6296
                                                                    4772
     2
                                        2
                                                                  21599
                        38100.0
                                                  25526
                        38100.0
                                        2
     3
                                                   5089
                                                                    3773
                                        2
     4
                        38100.0
                                                   4204
                                                                    3324
                            unemployment_rate_2012 affordable
                                                                  region_Far East
        ci90uball_2021
     0
                819527
                                                8.2
                                                               1
                                                                                 0
     1
                  7820
                                                7.1
                                                               1
                                                                                 0
     2
                  29453 ...
                                                7.7
                                                               1
                                                                                 0
                                               11.8
                                                                                 0
     3
                  6405
                                                               1
                  5084
                                                8.8
                                                               1
                                                                                 0
     4
        region_Great Lakes
                            region_Mid East region_New England region_Plains
     0
                          0
                                            0
                                                                 0
     1
                          0
                                            0
                                                                 0
                                                                                 0
     2
                          0
                                            0
                                                                 0
                                                                                 0
     3
                          0
                                            0
                                                                 0
                                                                                 0
     4
                          0
                                            0
                                                                 0
                                                                                 0
                                 region_Southeast
        region_Rocky Mountains
                                                    region_Southwest
     0
                                                  1
                                                                     0
     1
                              0
                                                  1
                                                                     0
     2
                              0
                                                  1
                                                                     0
     3
                              0
                                                  1
                                                                     0
     4
                              0
                                                                     0
                                                  1
```

[5 rows x 56 columns]

4.3 4.5 Normalize Numerical Values

```
[]: from sklearn.preprocessing import MinMaxScaler
                # Columns to apply MinMaxScaler
                numerical_features = ['costt4_a','tuitionfee_in','tuitionfee_out',_
                    -\unemployment_rate_2000', 'unemployment_rate_2001', 'unemployment_rate_2002', 'unemployment_rate_2002', 'unemployment_rate_2002', 'unemployment_rate_2001', 'unemployment_rat
                    _{\circlearrowleft} 'unemployment_rate_2004', 'unemployment_rate_2005',
                                                                                         'unemployment_rate_2006', u
                    → 'unemployment_rate_2007', 'unemployment_rate_2008', 'unemployment_rate_2009', 'unemployment_rate
                    → 'unemployment_rate_2013', 'unemployment_rate_2014', 'unemployment_rate_2015', 'unemployment_ra
                    'unemployment_rate_2019', 'unemployment_rate_2020', unemployment_rate_2020', unemployment_rate_2

¬'unemployment_rate_2021',

¬'unemployment_rate_2022', 'median_household_income_2021',
                    → 'avg_net_price_grants_scholarship_2020_2021', 'avg_net_price_grants_scholarship_2019_2020', '
                                                                                        'avg_net_price_grants_scholarship_2017_2018',u

¬'avg_net_price_grants_scholarship_2016_2017',

¬'avg_net_price_grants_scholarship_2015_2016',
                                                                                        'avg_net_price_grants_scholarship_2014_2015',

¬'avg_net_price_grants_scholarship_2013_2014',

¬'avg_net_price_grants_scholarship_2012_2013',

¬'avg_net_price_grants_scholarship_2011_2012',
                                                                                         'unemployment_rate_2012',]
                scaler = MinMaxScaler()
                # Store scaled values into new df
                scaled_df = encoded_df.copy()
                scaled_df[numerical_features] = scaler.

¬fit_transform(encoded_df[numerical_features])
                # Print scaled features
                scaled_df.head()
[]:
                         adm_rate_all sat_avg_all costt4_a tuitionfee_in tuitionfee_out \
                                      0.716006
                                                                                           954.0 0.278705
                                                                                                                                                                   0.262365
                                                                                                                                                                                                                          0.31364
                1
                                      0.716006
                                                                                           954.0 0.278705
                                                                                                                                                                   0.262365
                                                                                                                                                                                                                          0.31364
                2
                                      0.716006
                                                                                           954.0 0.278705
                                                                                                                                                                   0.262365
                                                                                                                                                                                                                          0.31364
                3
                                      0.716006
                                                                                           954.0 0.278705
                                                                                                                                                                   0.262365
                                                                                                                                                                                                                          0.31364
                                      0.716006
                                                                                           954.0 0.278705
                                                                                                                                                                   0.262365
                                                                                                                                                                                                                          0.31364
                         mn_earn_wne_indep1_p10 preddeg povall_2021 ci90lball_2021 \
```

```
0
                  0.193314
                                   2
                                         0.194248
                                                            782169
                                                               4772
1
                  0.193314
                                   2
                                         0.001526
2
                                   2
                                                              21599
                  0.193314
                                         0.006191
3
                  0.193314
                                   2
                                         0.001234
                                                               3773
                                   2
4
                  0.193314
                                         0.001019
                                                               3324
   ci90uball_2021 ... unemployment_rate_2012 affordable
                                                            region_Far East
0
           819527
                                      0.311404
                                                          1
                                                                             0
1
             7820
                                      0.263158
                                                          1
2
             29453 ...
                                      0.289474
                                                          1
                                                                             0
                                                                             0
3
             6405
                                      0.469298
                                                          1
             5084 ...
                                                          1
                                                                             0
4
                                      0.337719
   region_Great Lakes
                       region_Mid East region_New England region_Plains
0
                     0
                                       0
                                                            0
                     0
                                       0
                                                             0
                                                                             0
1
2
                     0
                                       0
                                                             0
                                                                             0
3
                     0
                                       0
                                                                             0
                                                             0
4
                     0
                                       0
                                                             0
                                                                             0
   region_Rocky Mountains region_Southeast region_Southwest
0
                                                                0
1
                         0
                                             1
                                                                0
2
                         0
                                                                0
                                             1
3
                         0
                                             1
                                                                0
                         0
                                             1
                                                                0
4
[5 rows x 56 columns]
```

[]: scaled_df.dtypes

[]:	adm_rate_all	float64
	sat_avg_all	float64
	costt4_a	float64
	tuitionfee_in	float64
	tuitionfee_out	float64
	mn_earn_wne_indep1_p10	float64
	preddeg	int64
	povall_2021	float64
	ci90lball_2021	int64
	ci90uball_2021	int64
	medhhinc_2021	float64
	ci90lbinc_2021	int64
	ci90ubinc_2021	int64
	unemployment_rate_2000	float64
	unemployment_rate_2001	float64
	unemployment_rate_2002	float64

unemployment_rate_2003	float64
unemployment_rate_2004	float64
unemployment_rate_2005	float64
unemployment_rate_2006	float64
unemployment_rate_2007	float64
unemployment_rate_2008	float64
unemployment_rate_2009	float64
unemployment_rate_2010	float64
unemployment_rate_2011	float64
unemployment_rate_2013	float64
unemployment_rate_2014	float64
unemployment_rate_2015	float64
unemployment_rate_2016	float64
unemployment_rate_2017	float64
unemployment_rate_2018	float64
unemployment_rate_2019	float64
unemployment_rate_2020	float64
unemployment_rate_2021	float64
unemployment_rate_2022	float64
median_household_income_2021	float64
<pre>avg_net_price_grants_scholarship_2020_2021</pre>	float64
<pre>avg_net_price_grants_scholarship_2019_2020</pre>	float64
<pre>avg_net_price_grants_scholarship_2018_2019</pre>	float64
<pre>avg_net_price_grants_scholarship_2017_2018</pre>	float64
<pre>avg_net_price_grants_scholarship_2016_2017</pre>	float64
<pre>avg_net_price_grants_scholarship_2015_2016</pre>	float64
<pre>avg_net_price_grants_scholarship_2014_2015</pre>	float64
<pre>avg_net_price_grants_scholarship_2013_2014</pre>	float64
<pre>avg_net_price_grants_scholarship_2012_2013</pre>	float64
<pre>avg_net_price_grants_scholarship_2011_2012</pre>	float64
unemployment_rate_2012	float64
affordable	int64
region_Far East	uint8
region_Great Lakes	uint8
region_Mid East	uint8
region_New England	uint8
region_Plains	uint8
region_Rocky Mountains	uint8
region_Southeast	uint8
region_Southwest	uint8
dtype: object	

5 5. Modeling

```
[]: from sklearn.model_selection import train_test_split
     # Define features and target variable
     X = scaled df.drop(columns=['affordable'])
     y = scaled df['affordable']
     # Split the data into train and temporary sets (70% train, 30% temp)
     X_train, X_temp, y_train, y_temp = train_test_split(X, y, test_size=0.3,_
      →random_state=42)
     # Split the temporary set into validation and test sets (50% for validation,
     450\% for test)
     X_valid, X_test, y_valid, y_test = train_test_split(X_temp, y_temp, test_size=0.
     →5, random_state=42)
[]: # Save data to CSV files
     train df = pd.concat([y train, X train], axis=1)
     valid_df = pd.concat([y_valid, X_valid], axis=1)
     test df = pd.concat([y test, X test], axis=1)
     # Save the combined DataFrame to a CSV file
     train_df.to_csv(os.path.join('data', 'train_combined.csv'), index=False)
     X_train.to_csv(os.path.join('data', 'train_features.csv'), index=False)
     y_train.to_csv(os.path.join('data', 'train_labels.csv'), index=False,__
      ⇔header=False)
     valid_df.to_csv(os.path.join('data', 'valid_combined.csv'), index=False)
     X_valid.to_csv(os.path.join('data', 'valid_features.csv'), index=False)
     y_valid.to_csv(os.path.join('data', 'valid_labels.csv'), index=False,__
      ⊸header=False)
     test_df.to_csv(os.path.join('data', 'test_combined.csv'), index=False)
     X_test.to_csv(os.path.join('data', 'test_features.csv'), index=False)
     y_test.to_csv(os.path.join('data', 'test_labels.csv'), index=False,__
      ⇔header=False)
     # Create a subset based on correlated attributes
     columns_to_include = ['affordable', 'sat_avg_all', 'costt4_a',_
     ⇔'avg_net_price_grants_scholarship_2020_2021',
                           'avg_net_price_grants_scholarship_2019_2020']
     train_subset = train_df[columns_to_include].copy()
     valid_subset = valid_df[columns_to_include].copy()
     test_subset = test_df[columns_to_include].copy()
```

```
train_subset.to_csv(os.path.join('data', 'train_subset.csv'), index=False,__
      ⇔header=False)
    valid_subset.to_csv(os.path.join('data', 'valid_subset.csv'), index=False,__
      ⇔header=False)
    test_subset.to_csv(os.path.join('data', 'test_subset.csv'), index=False,__
      ⇔header=False)
    X_test_subset = test_subset.drop(test_subset.columns[0], axis=1)
    X_test_subset.to_csv(os.path.join('data', 'X_test_subset.csv'), index=False,__
      ⇔header=False)
[]: # Print the shapes of the resulting sets
    print("Train set shape:", X_train.shape)
    print("Train set shape:", y_train.shape)
    print("Validation set shape:", X valid.shape)
    print("Validation set shape:", y_valid.shape)
    print("Test set shape:", X_test.shape)
    print("Test set shape:", y_test.shape)
    Train set shape: (22997, 55)
    Train set shape: (22997,)
    Validation set shape: (4928, 55)
    Validation set shape: (4928,)
    Test set shape: (4928, 55)
    Test set shape: (4928,)
[]: # Private bucket for JVo
    import boto3
    # Create a private S3 bucket
    s3 = boto3.resource('s3')
    bucket_name = "collegeaffordability508"
    bucket = s3.create_bucket(Bucket=bucket_name, ACL='private')
     # Upload CSV files to S3
    s3_client = boto3.client('s3')
    s3_private_path = "s3://{}/".format(bucket_name)
[]: # Upload train data
    s3_client.upload_file('data/train_subset.csv', bucket_name, 'train/train_subset.
    s3_client.upload_file('data/train_combined.csv', bucket_name, 'train/
     s3_client.upload_file('data/train_features.csv', bucket_name, 'train/
     ⇔train_features.csv')
```

```
s3_client.upload_file('data/train_labels.csv', bucket_name, 'train/train_labels.
 ⇔csv¹)
# Upload validation data
s3_client.upload_file('data/valid_subset.csv', bucket_name, 'validation/
⇔valid subset.csv')
s3_client.upload_file('data/valid_combined.csv', bucket_name, 'validation/
 ⇔valid_combined.csv')
s3_client.upload_file('data/valid_features.csv', bucket_name, 'validation/
⇔valid features.csv')
s3_client.upload_file('data/valid_labels.csv', bucket_name, 'validation/
⇔valid_labels.csv')
# Upload test data
s3_client.upload_file('data/test_subset.csv', bucket_name, 'test/test_subset.
s3_client.upload_file('data/X_test_subset.csv', bucket_name, 'test/

→X_test_subset.csv')
s3 client.upload file('data/test combined.csv', bucket name, 'test/
⇔test combined.csv')
s3_client.upload_file('data/test_features.csv', bucket_name, 'test/
 ⇔test_features.csv')
s3_client.upload_file('data/test_labels.csv', bucket_name, 'test/test_labels.
 ⇔csv')
```

5.0.1 5.1 Logistic Regression

```
[]: # Train an inital logistic regression model
log_reg_model = LogisticRegression()
log_reg_model.fit(X_train, y_train)

# Predictions on validation set
y_valid_pred = log_reg_model.predict(X_valid)

# Evaluate the model on validation set
valid_accuracy = accuracy_score(y_valid, y_valid_pred)
print("Validation Accuracy:", valid_accuracy.round(2))

# Predictions on test set
y_test_pred = log_reg_model.predict(X_test)

# Evaluate the model on test set
test_accuracy = accuracy_score(y_test, y_test_pred)
print("Test Accuracy:", test_accuracy.round(2))

# Print classification report
```

```
print("\nClassification Report:")
print(classification_report(y_valid, y_valid_pred))
```

Validation Accuracy: 0.76

Test Accuracy: 0.76

Classification Report:

	precision	recall	f1-score	support
0	0.33	0.00	0.00	1199
1	0.76	1.00	0.86	3729
accuracy			0.76	4928
macro avg	0.55	0.50	0.43	4928
weighted avg	0.65	0.76	0.65	4928

5.0.2 5.2 XG Boost

```
sagemaker.config INFO - Not applying SDK defaults from location:
/etc/xdg/sagemaker/config.yaml
sagemaker.config INFO - Not applying SDK defaults from location:
/root/.config/sagemaker/config.yaml
Success - the MySageMakerInstance is in the us-east-1 region. You will use the
811284229777.dkr.ecr.us-east-1.amazonaws.com/xgboost:latest container for your
SageMaker endpoint.
```

```
[]: # Load the data from S3
bucket_name = 'collegeaffordability508'
s3_client = boto3.client('s3')
```

```
train_data_uri = f's3://{bucket_name}/train_train_subset.csv'
     validation_data_uri = f's3://{bucket_name}/validation/valid_subset.csv'
     sess = sagemaker.Session()
     # Retrieve the XGBoost container image
     xgboost_container = retrieve(region=boto3.Session().region_name,__

¬framework='xgboost', version='latest')
     # Define the estimator
     xgb = sagemaker.estimator.Estimator(xgboost_container,
                                          role,
                                          instance_count=1,
                                          instance_type='ml.m5.xlarge',
                                          output_path=f's3://{bucket_name}/output',
                                          sagemaker_session=sess)
     # Parse in the hyperparameters
     xgb.set_hyperparameters(max_depth=5,
                             eta=0.2,
                             gamma=4,
                             min_child_weight=6,
                             subsample=0.8,
                             silent=0,
                             objective='binary:logistic',
                             num_round=100)
    sagemaker.config INFO - Not applying SDK defaults from location:
    /etc/xdg/sagemaker/config.yaml
    sagemaker.config INFO - Not applying SDK defaults from location:
    /root/.config/sagemaker/config.yaml
[]: # Define data channels for training
     train_input = TrainingInput(train_data_uri, content_type='text/csv')
     validation input = TrainingInput(validation data_uri, content_type='text/csv')
     # Train the XGBoost model
     xgb.fit({'train': train_input, 'validation': validation_input})
    INFO:sagemaker:Creating training-job with name: xgboost-2024-04-14-17-06-28-654
    2024-04-14 17:06:28 Starting - Starting the training job...
    2024-04-14 17:06:44 Starting - Preparing the instances for training...
    2024-04-14 17:07:15 Downloading - Downloading input data...
    2024\text{-}04\text{-}14 17\text{:}07\text{:}40 Downloading - Downloading the training image...
    2024-04-14 17:08:25 Training - Training image download completed. Training in
    progress.
    2024-04-14 17:08:25 Uploading - Uploading generated training
```

```
modelArguments: train
[2024-04-14:17:08:18:INFO] Running standalone xgboost training.
[2024-04-14:17:08:18:INFO] File size need to be processed in the node:
1.78mb. Available memory size in the node: 7994.32mb
[2024-04-14:17:08:18:INFO] Determined delimiter of CSV input is ','
[17:08:18] S3DistributionType set as FullyReplicated
[17:08:18] 22997x4 matrix with 91988 entries loaded from
/opt/ml/input/data/train?format=csv&label_column=0&delimiter=,
[2024-04-14:17:08:18:INFO] Determined delimiter of CSV input is ','
[17:08:18] S3DistributionType set as FullyReplicated
[17:08:18] 4928x4 matrix with 19712 entries loaded from
/opt/ml/input/data/validation?format=csv&label column=0&delimiter=,
[17:08:18] src/tree/updater prune.cc:74: tree pruning end, 1 roots, 36
extra nodes, 0 pruned nodes, max_depth=5
[0]#011train-error:0.069922#011validation-error:0.06737
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 38
extra nodes, 4 pruned nodes, max_depth=5
[1]#011train-error:0.054442#011validation-error:0.050122
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 46
extra nodes, 0 pruned nodes, max_depth=5
[2]#011train-error:0.042875#011validation-error:0.041599
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 44
extra nodes, 4 pruned nodes, max_depth=5
[3]#011train-error:0.035918#011validation-error:0.033279
[17:08:18] src/tree/updater prune.cc:74: tree pruning end, 1 roots, 38
extra nodes, 2 pruned nodes, max depth=5
[4]#011train-error:0.032222#011validation-error:0.026583
[17:08:18] src/tree/updater prune.cc:74: tree pruning end, 1 roots, 38
extra nodes, 0 pruned nodes, max_depth=5
[5]#011train-error:0.034135#011validation-error:0.029627
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 34
extra nodes, 2 pruned nodes, max_depth=5
[6]#011train-error:0.031526#011validation-error:0.026989
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 48
extra nodes, 4 pruned nodes, max_depth=5
[7]#011train-error:0.028221#011validation-error:0.024148
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 34
extra nodes, 0 pruned nodes, max depth=5
[8]#011train-error:0.023525#011validation-error:0.018466
```

```
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 34
extra nodes, 4 pruned nodes, max_depth=5
[9]#011train-error:0.023525#011validation-error:0.018466
[17:08:18] src/tree/updater prune.cc:74: tree pruning end, 1 roots, 42
extra nodes, 2 pruned nodes, max_depth=5
[10]#011train-error:0.023525#011validation-error:0.018466
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 42
extra nodes, 4 pruned nodes, max depth=5
[11]#011train-error:0.017437#011validation-error:0.012784
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 40
extra nodes, 4 pruned nodes, max_depth=5
[12]#011train-error:0.015741#011validation-error:0.011161
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 30
extra nodes, 4 pruned nodes, max_depth=5
[13]#011train-error:0.013263#011validation-error:0.00832
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 30
extra nodes, 4 pruned nodes, max_depth=5
[14]#011train-error:0.010132#011validation-error:0.006494
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 28
extra nodes, 8 pruned nodes, max depth=5
[15]#011train-error:0.010132#011validation-error:0.006494
[17:08:18] src/tree/updater prune.cc:74: tree pruning end, 1 roots, 34
extra nodes, 0 pruned nodes, max_depth=5
[16]#011train-error:0.010132#011validation-error:0.006494
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 32
extra nodes, 2 pruned nodes, max_depth=5
[17]#011train-error:0.008131#011validation-error:0.00487
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 24
extra nodes, 0 pruned nodes, max depth=5
[18] #011train-error:0.007175#011validation-error:0.004464
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 32
extra nodes, 2 pruned nodes, max_depth=5
[19]#011train-error:0.007175#011validation-error:0.004464
[17:08:18] src/tree/updater prune.cc:74: tree pruning end, 1 roots, 28
extra nodes, 4 pruned nodes, max_depth=5
[20]#011train-error:0.008131#011validation-error:0.00487
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 38
extra nodes, 4 pruned nodes, max_depth=5
[21]#011train-error:0.00587#011validation-error:0.003653
```

```
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 24
extra nodes, 0 pruned nodes, max_depth=5
[22]#011train-error:0.007175#011validation-error:0.004464
[17:08:18] src/tree/updater prune.cc:74: tree pruning end, 1 roots, 36
extra nodes, 0 pruned nodes, max_depth=5
[23]#011train-error:0.003305#011validation-error:0.002232
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 18
extra nodes, 4 pruned nodes, max depth=5
[24]#011train-error:0.002131#011validation-error:0.001623
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 28
extra nodes, 0 pruned nodes, max_depth=5
[25]#011train-error:0.002131#011validation-error:0.001623
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 26
extra nodes, 2 pruned nodes, max_depth=5
[26]#011train-error:0.002131#011validation-error:0.001623
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 20
extra nodes, 0 pruned nodes, max_depth=5
[27]#011train-error:0.002131#011validation-error:0.001623
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 26
extra nodes, 0 pruned nodes, max depth=5
[28] #011train-error:0.002131#011validation-error:0.001623
[17:08:18] src/tree/updater prune.cc:74: tree pruning end, 1 roots, 18
extra nodes, 4 pruned nodes, max_depth=5
[29]#011train-error:0.001913#011validation-error:0.00142
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 24
extra nodes, 2 pruned nodes, max_depth=5
[30]#011train-error:0.00187#011validation-error:0.000812
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 20
extra nodes, 4 pruned nodes, max depth=5
[31]#011train-error:0.00187#011validation-error:0.000812
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 12
extra nodes, 4 pruned nodes, max_depth=5
[32]#011train-error:0.001522#011validation-error:0.000609
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 28
extra nodes, 2 pruned nodes, max_depth=5
[33] #011train-error:0.001174#011validation-error:0.000406
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 16
extra nodes, 4 pruned nodes, max_depth=5
[34]#011train-error:0.001174#011validation-error:0.000406
```

```
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 14
extra nodes, 4 pruned nodes, max_depth=5
[35]#011train-error:0.001522#011validation-error:0.000609
[17:08:18] src/tree/updater prune.cc:74: tree pruning end, 1 roots, 14
extra nodes, 4 pruned nodes, max_depth=5
[36] #011train-error:0.001522#011validation-error:0.000609
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 42
extra nodes, 4 pruned nodes, max depth=5
[37]#011train-error:0.000957#011validation-error:0.000203
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 32
extra nodes, 2 pruned nodes, max_depth=5
[38]#011train-error:0.000957#011validation-error:0.000203
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 12
extra nodes, 6 pruned nodes, max depth=5
[39]#011train-error:0.000609#011validation-error:0
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 14
extra nodes, 4 pruned nodes, max_depth=5
[40]#011train-error:0.000609#011validation-error:0
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 26
extra nodes, 4 pruned nodes, max_depth=5
[41]#011train-error:0.000609#011validation-error:0
[17:08:18] src/tree/updater prune.cc:74: tree pruning end, 1 roots, 22
extra nodes, 2 pruned nodes, max_depth=5
[42]#011train-error:0.000609#011validation-error:0
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 24
extra nodes, 8 pruned nodes, max depth=5
[43]#011train-error:0.000435#011validation-error:0
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 22
extra nodes, 2 pruned nodes, max depth=5
[44]#011train-error:0.000435#011validation-error:0
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 14
extra nodes, 6 pruned nodes, max_depth=5
[45]#011train-error:0.000435#011validation-error:0
[17:08:18] src/tree/updater prune.cc:74: tree pruning end, 1 roots, 16
extra nodes, 2 pruned nodes, max_depth=5
[46]#011train-error:0.000304#011validation-error:0
[17:08:18] src/tree/updater prune.cc:74: tree pruning end, 1 roots, 18
extra nodes, 4 pruned nodes, max_depth=5
[47]#011train-error:0#011validation-error:0
```

```
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 14
extra nodes, 2 pruned nodes, max_depth=5
[48] #011train-error:0#011validation-error:0
[17:08:18] src/tree/updater prune.cc:74: tree pruning end, 1 roots, 16
extra nodes, 8 pruned nodes, max_depth=5
[49]#011train-error:0.000304#011validation-error:0
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 20
extra nodes, 6 pruned nodes, max depth=5
[50]#011train-error:0#011validation-error:0
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 22
extra nodes, 8 pruned nodes, max_depth=5
[51]#011train-error:0#011validation-error:0
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 10
extra nodes, 10 pruned nodes, max depth=5
[52]#011train-error:0#011validation-error:0
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 12
extra nodes, 6 pruned nodes, max_depth=5
[53]#011train-error:0#011validation-error:0
[17:08:18] src/tree/updater prune.cc:74: tree pruning end, 1 roots, 14
extra nodes, 2 pruned nodes, max_depth=5
[54] #011train-error:0#011validation-error:0
[17:08:18] src/tree/updater prune.cc:74: tree pruning end, 1 roots, 22
extra nodes, 2 pruned nodes, max_depth=5
[55]#011train-error:0#011validation-error:0
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 16
extra nodes, 2 pruned nodes, max depth=5
[56] #011train-error:0#011validation-error:0
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 10
extra nodes, 4 pruned nodes, max depth=5
[57]#011train-error:0#011validation-error:0
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 12
extra nodes, 4 pruned nodes, max_depth=5
[58] #011train-error:0#011validation-error:0
[17:08:18] src/tree/updater prune.cc:74: tree pruning end, 1 roots, 22
extra nodes, 6 pruned nodes, max_depth=5
[59]#011train-error:0#011validation-error:0
[17:08:18] src/tree/updater prune.cc:74: tree pruning end, 1 roots, 6 extra
nodes, 14 pruned nodes, max_depth=3
[60]#011train-error:0#011validation-error:0
```

```
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 16
extra nodes, 8 pruned nodes, max_depth=5
[61]#011train-error:0#011validation-error:0
[17:08:18] src/tree/updater prune.cc:74: tree pruning end, 1 roots, 12
extra nodes, 2 pruned nodes, max_depth=5
[62] #011train-error:0#011validation-error:0
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 6 extra
nodes, 6 pruned nodes, max depth=3
[63]#011train-error:0#011validation-error:0
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 16
extra nodes, 8 pruned nodes, max_depth=4
[64]#011train-error:0#011validation-error:0
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 12
extra nodes, 4 pruned nodes, max_depth=5
[65]#011train-error:0#011validation-error:0
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 20
extra nodes, 2 pruned nodes, max_depth=5
[66]#011train-error:0#011validation-error:0
[17:08:18] src/tree/updater prune.cc:74: tree pruning end, 1 roots, 10
extra nodes, 12 pruned nodes, max depth=4
[67]#011train-error:0#011validation-error:0
[17:08:18] src/tree/updater prune.cc:74: tree pruning end, 1 roots, 10
extra nodes, 8 pruned nodes, max_depth=5
[68]#011train-error:0#011validation-error:0
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 0 extra
nodes, 12 pruned nodes, max depth=0
[69]#011train-error:0#011validation-error:0
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 10
extra nodes, 6 pruned nodes, max depth=4
[70]#011train-error:0#011validation-error:0
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 10
extra nodes, 6 pruned nodes, max_depth=5
[71]#011train-error:0#011validation-error:0
[17:08:18] src/tree/updater prune.cc:74: tree pruning end, 1 roots, 10
extra nodes, 6 pruned nodes, max_depth=5
[72] #011train-error:0#011validation-error:0
[17:08:18] src/tree/updater prune.cc:74: tree pruning end, 1 roots, 20
extra nodes, 8 pruned nodes, max_depth=5
[73]#011train-error:0#011validation-error:0
```

```
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 0 extra
nodes, 20 pruned nodes, max_depth=0
[74]#011train-error:0#011validation-error:0
[17:08:18] src/tree/updater prune.cc:74: tree pruning end, 1 roots, 6 extra
nodes, 14 pruned nodes, max_depth=3
[75]#011train-error:0#011validation-error:0
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 10
extra nodes, 2 pruned nodes, max depth=5
[76]#011train-error:0#011validation-error:0
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 0 extra
nodes, 12 pruned nodes, max_depth=0
[77]#011train-error:0#011validation-error:0
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 18
extra nodes, 8 pruned nodes, max_depth=5
[78]#011train-error:0#011validation-error:0
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 0 extra
nodes, 12 pruned nodes, max_depth=0
[79]#011train-error:0#011validation-error:0
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 10
extra nodes, 4 pruned nodes, max_depth=5
[80]#011train-error:0#011validation-error:0
[17:08:18] src/tree/updater prune.cc:74: tree pruning end, 1 roots, 8 extra
nodes, 14 pruned nodes, max_depth=4
[81] #011train-error:0#011validation-error:0
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 14
extra nodes, 12 pruned nodes, max depth=5
[82] #011train-error: 0#011validation-error: 0
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 10
extra nodes, 14 pruned nodes, max depth=5
[83]#011train-error:0#011validation-error:0
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 0 extra
nodes, 18 pruned nodes, max_depth=0
[84]#011train-error:0#011validation-error:0
[17:08:18] src/tree/updater prune.cc:74: tree pruning end, 1 roots, 0 extra
nodes, 20 pruned nodes, max_depth=0
[85]#011train-error:0#011validation-error:0
[17:08:18] src/tree/updater prune.cc:74: tree pruning end, 1 roots, 10
extra nodes, 10 pruned nodes, max_depth=5
[86]#011train-error:0#011validation-error:0
```

```
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 0 extra
nodes, 20 pruned nodes, max_depth=0
[87]#011train-error:0#011validation-error:0
[17:08:18] src/tree/updater prune.cc:74: tree pruning end, 1 roots, 10
extra nodes, 8 pruned nodes, max_depth=5
[88] #011train-error:0#011validation-error:0
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 0 extra
nodes, 12 pruned nodes, max depth=0
[89]#011train-error:0#011validation-error:0
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 0 extra
nodes, 14 pruned nodes, max_depth=0
[90]#011train-error:0#011validation-error:0
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 10
extra nodes, 8 pruned nodes, max depth=5
[91]#011train-error:0#011validation-error:0
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 6 extra
nodes, 6 pruned nodes, max_depth=3
[92]#011train-error:0#011validation-error:0
[17:08:18] src/tree/updater prune.cc:74: tree pruning end, 1 roots, 0 extra
nodes, 10 pruned nodes, max_depth=0
[93]#011train-error:0#011validation-error:0
[17:08:18] src/tree/updater prune.cc:74: tree pruning end, 1 roots, 12
extra nodes, 4 pruned nodes, max_depth=5
[94]#011train-error:0#011validation-error:0
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 0 extra
nodes, 16 pruned nodes, max depth=0
[95]#011train-error:0#011validation-error:0
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 0 extra
nodes, 20 pruned nodes, max depth=0
[96]#011train-error:0#011validation-error:0
[17:08:18] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 0 extra
nodes, 10 pruned nodes, max_depth=0
[97]#011train-error:0#011validation-error:0
[17:08:18] src/tree/updater prune.cc:74: tree pruning end, 1 roots, 0 extra
nodes, 18 pruned nodes, max_depth=0
[98]#011train-error:0#011validation-error:0
[17:08:18] src/tree/updater prune.cc:74: tree pruning end, 1 roots, 6 extra
nodes, 16 pruned nodes, max_depth=3
[99]#011train-error:0#011validation-error:0
```

```
Billable seconds: 82
[]: from sagemaker.serializers import CSVSerializer
     from sklearn.metrics import accuracy_score, classification_report
     # Load the test labels from S3
     test_labels_uri = f's3://{bucket_name}/test/test_labels.csv'
     test_labels = pd.read_csv(test_labels_uri)
     # Assuming test_labels contains the ground truth labels
     # Define the test data URI
     test_data_uri = f's3://{bucket_name}/test/X_test_subset.csv'
     # Read the test data from S3
     test_data = pd.read_csv(test_data_uri)
     # Convert the test data to CSV format
     csv_data = test_data.to_csv(index=False, header=False).rstrip('\n')
     # Deploy the model
     xgb_predictor = xgb.deploy(initial_instance_count=1, instance_type='ml.m5.
      →large', serializer=CSVSerializer())
     # Evaluate the model on the test data
     test_predictions = xgb_predictor.predict(csv_data).decode('utf-8')
     # Convert predictions to DataFrame
     predicted_labels = pd.DataFrame([int(float(x)) for x in test_predictions.
      ⇔split(',')])
     # Calculate accuracy score
     accuracy = accuracy_score(test_labels, predicted_labels)
     # Generate classification report
     classification_rep = classification_report(test_labels, predicted_labels)
     print("Accuracy Score:", accuracy)
     print("Classification Report:\n", classification_rep)
    INFO:sagemaker:Creating model with name: xgboost-2024-04-14-17-56-14-238
    INFO:sagemaker:Creating endpoint-config with name
    xgboost-2024-04-14-17-56-14-238
    INFO:sagemaker:Creating endpoint with name xgboost-2024-04-14-17-56-14-238
    ----!Accuracy Score: 0.23807590826060482
```

2024-04-14 17:08:37 Completed - Training job completed

Training seconds: 82

Classification Report:

	precision	recall	f1-score	support
0	0.24	1.00	0.38	1173
1	0.00	0.00	0.00	3754
accuracy			0.24	4927
macro avg	0.12	0.50	0.19	4927
weighted avg	0.06	0.24	0.09	4927

/opt/conda/lib/python3.8/site-packages/sklearn/metrics/_classification.py:1471: UndefinedMetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero_division` parameter to control this behavior.

```
_warn_prf(average, modifier, msg_start, len(result))
```

/opt/conda/lib/python3.8/site-packages/sklearn/metrics/_classification.py:1471: UndefinedMetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero_division` parameter to control this behavior.

```
_warn_prf(average, modifier, msg_start, len(result))
```

/opt/conda/lib/python3.8/site-packages/sklearn/metrics/_classification.py:1471: UndefinedMetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero_division` parameter to control this behavior.

```
_warn_prf(average, modifier, msg_start, len(result))
```

Try Improving the Model with Randomly Selected Hyperparameters

```
role,
                                    instance_count=1,
                                    instance_type='ml.m5.xlarge',
                                    output_path=f's3://{bucket_name}/output',
                                    sagemaker_session=sess)
# Define hyperparameter ranges for random search
hyperparameter_ranges = {
    "max depth": IntegerParameter(3, 10),
    "eta": ContinuousParameter(0.01, 0.5),
    "gamma": ContinuousParameter(0, 10),
    "min_child_weight": IntegerParameter(1, 10),
    "subsample": ContinuousParameter(0.5, 1),
    "num_round": IntegerParameter(50, 200)
}
# Define the objective metric for hyperparameter optimization
objective_metric_name = "validation:auc"
# Define the hyperparameter tuner
hyperparameter_tuner = HyperparameterTuner(estimator=xgb,
 ⇔objective_metric_name=objective_metric_name,
 hyperparameter_ranges=hyperparameter_ranges,
                                           strategy="Random",
                                           max jobs=10,
                                           max_parallel_jobs=2)
```

```
sagemaker.config INFO - Not applying SDK defaults from location:
/etc/xdg/sagemaker/config.yaml
sagemaker.config INFO - Not applying SDK defaults from location:
/root/.config/sagemaker/config.yaml
```

INFO:sagemaker.image_uris:Ignoring unnecessary instance type: None.

```
[]: # Define data channels for training
train_input = TrainingInput(train_data_uri, content_type='text/csv')
validation_input = TrainingInput(validation_data_uri, content_type='text/csv')

# Train the model with hyperparameter tuning
hyperparameter_tuner.fit({"train": train_input, "validation": validation_input})
```

WARNING:sagemaker.estimator:No finished training job found associated with this estimator. Please make sure this estimator is only used for building workflow config

WARNING:sagemaker.estimator:No finished training job found associated with this estimator. Please make sure this estimator is only used for building workflow

```
config
    INFO:sagemaker:Creating hyperparameter tuning job with name: xgboost-240414-1801
    ...!
[]: # Define the test data URI
    test data uri = f's3://{bucket name}/test/X test subset.csv'
    test_labels_uri = f's3://{bucket_name}/test/test_labels.csv'
    # Load the test data and labels from S3
    test_data = pd.read_csv(test_data_uri)
    test_labels = pd.read_csv(test_labels_uri)
    # Deploy the best model found during hyperparameter tuning
    xgb_predictor = hyperparameter_tuner.deploy(initial_instance_count=1,__
      # Convert the test data to CSV format
    csv_data = test_data.to_csv(index=False, header=False).rstrip('\n')
    # Evaluate the model on the test data
    test_predictions = xgb_predictor.predict(csv_data).decode('utf-8')
    # Convert predictions to DataFrame
    predicted_labels = pd.DataFrame([int(float(x)) for x in test_predictions.
      ⇔split(',')])
    # Calculate accuracy score
    accuracy = accuracy_score(test_labels, predicted_labels)
    # Generate classification report
    classification rep = classification report(test_labels, predicted_labels)
    print("Accuracy Score:", accuracy)
    print("Classification Report:\n", classification_rep)
    2024-04-14 18:04:20 Starting - Preparing the instances for training
    2024-04-14 18:04:20 Downloading - Downloading the training image
    2024-04-14 18:04:20 Training - Training image download completed. Training in
    progress.
    2024-04-14 18:04:20 Uploading - Uploading generated training model
    2024-04-14 18:04:20 Completed - Resource reused by training job:
    xgboost-240414-1801-004-040277d0
    INFO:sagemaker:Creating model with name: xgboost-2024-04-14-19-13-11-797
```

INFO:sagemaker:Creating endpoint-config with name

```
xgboost-240414-1801-002-799b86a7
INFO:sagemaker:Creating endpoint with name xgboost-240414-1801-002-799b86a7
----!Accuracy Score: 0.3109397199106962
Classification Report:
               precision
                           recall f1-score
                                               support
           0
                   0.26
                             1.00
                                       0.41
                                                 1173
           1
                   1.00
                             0.10
                                       0.17
                                                 3754
                                       0.31
                                                 4927
   accuracy
  macro avg
                   0.63
                             0.55
                                       0.29
                                                 4927
weighted avg
                   0.82
                             0.31
                                       0.23
                                                 4927
```

```
[]:  # Delete the endpoint to keep costs minimal 
xgb_predictor.delete_endpoint()
```

INFO: sagemaker: Deleting endpoint configuration with name:

xgboost-240414-1801-002-799b86a7

INFO:sagemaker:Deleting endpoint with name: xgboost-240414-1801-002-799b86a7

5.0.3 5.3 Random Forrest

```
[]: from sklearn.ensemble import RandomForestClassifier
     rf_model = RandomForestClassifier(n_estimators = 500,
                                 max_depth = 4,
                                 max_features = 3,
                                 bootstrap = True,
                                 random_state = 508)
     rf_model.fit(X_train, y_train)
     y_valid_pred = rf_model.predict(X_valid)
     # Evaluate the model on validation set
     valid_accuracy = accuracy_score(y_valid, y_valid_pred)
     print("Validation Accuracy:", valid_accuracy.round(2))
     # Predictions on test set
     y_test_pred = rf_model.predict(X_test)
     # Evaluate the model on test set
     test_accuracy = accuracy_score(y_test, y_test_pred)
     print("Test Accuracy:", test_accuracy.round(2))
     # Print classification report
```

```
print("\nClassification Report:")
print(classification_report(y_valid, y_valid_pred))
```

Validation Accuracy: 0.95 Test Accuracy: 0.95

Classification Report:

	precision	recall	f1-score	support
0	1.00	0.79	0.89	1199
1	0.94	1.00	0.97	3729
accuracy			0.95	4928
macro avg	0.97	0.90	0.93	4928
weighted avg	0.95	0.95	0.95	4928

Release Sagemaker Resources

```
try {
    Jupyter.notebook.save_checkpoint();
    Jupyter.notebook.session.delete();
}
catch(err) {
    // NoOp
}
```

<IPython.core.display.Javascript object>

<IPython.core.display.HTML object>

Reading package lists... Done Building dependency tree... Done Reading state information... Done

The following additional packages will be installed:

dvisvgm fonts-droid-fallback fonts-lato fonts-lmodern fonts-noto-mono fonts-texgyre fonts-urw-base35 libapache-pom-java libcommons-logging-java libcommons-parent-java libfontbox-java libfontenc1 libgs9 libgs9-common libidn12 libijs-0.35 libjbig2dec0 libkpathsea6 libpdfbox-java libptexenc1 libruby3.0 libsynctex2 libteckit0 libtexlua53 libtexluajit2 libwoff1 libzzip-0-13 lmodern poppler-data preview-latex-style rake ruby ruby-net-telnet ruby-rubygems ruby-webrick ruby-xmlrpc ruby3.0 rubygems-integration tlutils teckit tex-common tex-gyre texlive-base texlive-binaries texlive-latex-base texlive-latex-extra texlive-latex-recommended texlive-pictures tipa xfonts-encodings xfonts-utils

Suggested packages:

fonts-noto fonts-freefont-otf | fonts-freefont-ttf libavalon-framework-java libcommons-logging-java-doc libexcalibur-logkit-java liblog4j1.2-java poppler-utils ghostscript fonts-japanese-mincho | fonts-ipafont-mincho fonts-japanese-gothic | fonts-ipafont-gothic fonts-arphic-ukai fonts-arphic-uming fonts-nanum ri ruby-dev bundler debhelper gv | postscript-viewer perl-tk xpdf | pdf-viewer xzdec texlive-fonts-recommended-doc texlive-latex-base-doc python3-pygments icc-profiles libfile-which-perl libspreadsheet-parseexcel-perl texlive-latex-extra-doc texlive-latex-recommended-doc texlive-luatex texlive-pstricks dot2tex prerex texlive-pictures-doc vprerex default-jre-headless tipa-doc

The following NEW packages will be installed:

dvisvgm fonts-droid-fallback fonts-lato fonts-lmodern fonts-noto-mono fonts-texgyre fonts-urw-base35 libapache-pom-java libcommons-logging-java libcommons-parent-java libfontbox-java libfontenc1 libgs9 libgs9-common libidn12 libijs-0.35 libjbig2dec0 libkpathsea6 libpdfbox-java libptexenc1 libruby3.0 libsynctex2 libteckit0 libtexlua53 libtexluajit2 libwoff1 libzzip-0-13 lmodern poppler-data preview-latex-style rake ruby ruby-net-telnet ruby-rubygems ruby-webrick ruby-xmlrpc ruby3.0 rubygems-integration t1utils teckit tex-common tex-gyre texlive-base texlive-binaries texlive-fonts-recommended texlive-latex-base

```
texlive-latex-extra texlive-latex-recommended texlive-pictures
```

texlive-plain-generic texlive-xetex tipa xfonts-encodings xfonts-utils

O upgraded, 54 newly installed, O to remove and 45 not upgraded.

Need to get 182 MB of archives.

After this operation, 571 MB of additional disk space will be used.

Get:1 http://archive.ubuntu.com/ubuntu jammy/main amd64 fonts-droid-fallback all 1:6.0.1r16-1.1build1 [1,805 kB]

Get:2 http://archive.ubuntu.com/ubuntu jammy/main amd64 fonts-lato all 2.0-2.1 [2,696 kB]

Get:3 http://archive.ubuntu.com/ubuntu jammy/main amd64 poppler-data all 0.4.11-1 [2,171 kB]

Get:4 http://archive.ubuntu.com/ubuntu jammy/universe amd64 tex-common all 6.17
[33.7 kB]

Get:5 http://archive.ubuntu.com/ubuntu jammy/main amd64 fonts-urw-base35 all 20200910-1 [6,367 kB]

Get:6 http://archive.ubuntu.com/ubuntu jammy-updates/main amd64 libgs9-common
all 9.55.0~dfsg1-Oubuntu5.6 [751 kB]

Get:7 http://archive.ubuntu.com/ubuntu jammy-updates/main amd64 libidn12 amd64 1.38-4ubuntu1 [60.0 kB]

Get:8 http://archive.ubuntu.com/ubuntu jammy/main amd64 libijs-0.35 amd64 0.35-15build2 [16.5 kB]

Get:9 http://archive.ubuntu.com/ubuntu jammy/main amd64 libjbig2dec0 amd64 0.19-3build2 [64.7 kB]

Get:10 http://archive.ubuntu.com/ubuntu jammy-updates/main amd64 libgs9 amd64 9.55.0~dfsg1-Oubuntu5.6 [5,031 kB]

Get:11 http://archive.ubuntu.com/ubuntu jammy-updates/main amd64 libkpathsea6 amd64 2021.20210626.59705-1ubuntu0.2 [60.4 kB]

Get:12 http://archive.ubuntu.com/ubuntu jammy/main amd64 libwoff1 amd64
1.0.2-1build4 [45.2 kB]

Get:13 http://archive.ubuntu.com/ubuntu jammy/universe amd64 dvisvgm amd64 2.13.1-1 [1,221 kB]

Get:14 http://archive.ubuntu.com/ubuntu jammy/universe amd64 fonts-lmodern all 2.004.5-6.1 [4,532 kB]

Get:15 http://archive.ubuntu.com/ubuntu jammy/main amd64 fonts-noto-mono all 20201225-1build1 [397 kB]

Get:16 http://archive.ubuntu.com/ubuntu jammy/universe amd64 fonts-texgyre all 20180621-3.1 [10.2 MB]

Get:17 http://archive.ubuntu.com/ubuntu jammy/universe amd64 libapache-pom-java all 18-1 [4,720 B]

Get:18 http://archive.ubuntu.com/ubuntu jammy/universe amd64 libcommons-parent-java all 43-1 [10.8 kB]

Get:19 http://archive.ubuntu.com/ubuntu jammy/universe amd64 libcommons-logging-java all 1.2-2 [60.3 kB]

Get:20 http://archive.ubuntu.com/ubuntu jammy/main amd64 libfontenc1 amd64 1:1.1.4-1build3 [14.7 kB]

Get:21 http://archive.ubuntu.com/ubuntu jammy-updates/main amd64 libptexenc1 amd64 2021.20210626.59705-1ubuntu0.2 [39.1 kB]

Get:22 http://archive.ubuntu.com/ubuntu jammy/main amd64 rubygems-integration

```
all 1.18 [5,336 B]
```

Get:23 http://archive.ubuntu.com/ubuntu jammy-updates/main amd64 ruby3.0 amd64 3.0.2-7ubuntu2.4 [50.1 kB]

Get:24 http://archive.ubuntu.com/ubuntu jammy/main amd64 ruby-rubygems all 3.3.5-2 [228 kB]

Get:25 http://archive.ubuntu.com/ubuntu jammy/main amd64 ruby amd64 1:3.0~exp1
[5,100 B]

Get:26 http://archive.ubuntu.com/ubuntu jammy/main amd64 rake all 13.0.6-2 [61.7 kB]

Get:27 http://archive.ubuntu.com/ubuntu jammy/main amd64 ruby-net-telnet all 0.1.1-2 [12.6 kB]

Get:28 http://archive.ubuntu.com/ubuntu jammy/universe amd64 ruby-webrick all 1.7.0-3 [51.8 kB]

Get:29 http://archive.ubuntu.com/ubuntu jammy-updates/main amd64 ruby-xmlrpc all 0.3.2-1ubuntu0.1 [24.9 kB]

Get:30 http://archive.ubuntu.com/ubuntu jammy-updates/main amd64 libruby3.0 amd64 3.0.2-7ubuntu2.4 [5,113 kB]

Get:31 http://archive.ubuntu.com/ubuntu jammy-updates/main amd64 libsynctex2 amd64 2021.20210626.59705-1ubuntu0.2 [55.6 kB]

Get:32 http://archive.ubuntu.com/ubuntu jammy/universe amd64 libteckit0 amd64 2.5.11+ds1-1 [421 kB]

Get:33 http://archive.ubuntu.com/ubuntu jammy-updates/main amd64 libtexlua53 amd64 2021.20210626.59705-1ubuntu0.2 [120 kB]

Get:34 http://archive.ubuntu.com/ubuntu jammy-updates/main amd64 libtexluajit2 amd64 2021.20210626.59705-1ubuntu0.2 [267 kB]

Get:35 http://archive.ubuntu.com/ubuntu jammy/universe amd64 libzzip-0-13 amd64 0.13.72+dfsg.1-1.1 [27.0 kB]

Get:36 http://archive.ubuntu.com/ubuntu jammy/main amd64 xfonts-encodings all
1:1.0.5-Oubuntu2 [578 kB]

Get:37 http://archive.ubuntu.com/ubuntu jammy/main amd64 xfonts-utils amd64 1:7.7+6build2 [94.6 kB]

Get:38 http://archive.ubuntu.com/ubuntu jammy/universe amd64 lmodern all
2.004.5-6.1 [9,471 kB]

Get:39 http://archive.ubuntu.com/ubuntu jammy/universe amd64 preview-latex-style all 12.2-1ubuntu1 [185 kB]

Get:40 http://archive.ubuntu.com/ubuntu jammy/main amd64 t1utils amd64
1.41-4build2 [61.3 kB]

Get:41 http://archive.ubuntu.com/ubuntu jammy/universe amd64 teckit amd64
2.5.11+ds1-1 [699 kB]

Get:42 http://archive.ubuntu.com/ubuntu jammy/universe amd64 tex-gyre all 20180621-3.1 [6,209 kB]

Get:43 http://archive.ubuntu.com/ubuntu jammy-updates/universe amd64 texlive-binaries amd64 2021.20210626.59705-1ubuntu0.2 [9,860 kB]

Get:44 http://archive.ubuntu.com/ubuntu jammy/universe amd64 texlive-base all 2021.20220204-1 [21.0 MB]

Get:45 http://archive.ubuntu.com/ubuntu jammy/universe amd64 texlive-fonts-recommended all 2021.20220204-1 [4,972 kB]

Get:46 http://archive.ubuntu.com/ubuntu jammy/universe amd64 texlive-latex-base

```
all 2021.20220204-1 [1,128 kB]
Get:47 http://archive.ubuntu.com/ubuntu jammy/universe amd64 libfontbox-java all
1:1.8.16-2 [207 kB]
Get:48 http://archive.ubuntu.com/ubuntu jammy/universe amd64 libpdfbox-java all
1:1.8.16-2 [5,199 kB]
Get:49 http://archive.ubuntu.com/ubuntu jammy/universe amd64 texlive-latex-
recommended all 2021.20220204-1 [14.4 MB]
Get:50 http://archive.ubuntu.com/ubuntu jammy/universe amd64 texlive-pictures
all 2021.20220204-1 [8,720 kB]
Get:51 http://archive.ubuntu.com/ubuntu jammy/universe amd64 texlive-latex-extra
all 2021.20220204-1 [13.9 MB]
Get:52 http://archive.ubuntu.com/ubuntu jammy/universe amd64 texlive-plain-
generic all 2021.20220204-1 [27.5 MB]
Get:53 http://archive.ubuntu.com/ubuntu jammy/universe amd64 tipa all 2:1.3-21
[2,967 \text{ kB}]
Get:54 http://archive.ubuntu.com/ubuntu jammy/universe amd64 texlive-xetex all
2021.20220204-1 [12.4 MB]
Fetched 182 MB in 15s (12.2 MB/s)
debconf: unable to initialize frontend: Dialog
debconf: (No usable dialog-like program is installed, so the dialog based
frontend cannot be used. at /usr/share/perl5/Debconf/FrontEnd/Dialog.pm line 78,
<> line 54.)
debconf: falling back to frontend: Readline
debconf: unable to initialize frontend: Readline
debconf: (This frontend requires a controlling tty.)
debconf: falling back to frontend: Teletype
dpkg-preconfigure: unable to re-open stdin:
Selecting previously unselected package fonts-droid-fallback.
(Reading database ... 121752 files and directories currently installed.)
Preparing to unpack .../00-fonts-droid-fallback_1%3a6.0.1r16-1.1build1_all.deb
Unpacking fonts-droid-fallback (1:6.0.1r16-1.1build1) ...
Selecting previously unselected package fonts-lato.
Preparing to unpack .../01-fonts-lato_2.0-2.1_all.deb ...
Unpacking fonts-lato (2.0-2.1) ...
Selecting previously unselected package poppler-data.
Preparing to unpack .../02-poppler-data 0.4.11-1 all.deb ...
Unpacking poppler-data (0.4.11-1) ...
Selecting previously unselected package tex-common.
Preparing to unpack .../03-tex-common_6.17_all.deb ...
Unpacking tex-common (6.17) ...
Selecting previously unselected package fonts-urw-base35.
Preparing to unpack .../04-fonts-urw-base35_20200910-1_all.deb ...
Unpacking fonts-urw-base35 (20200910-1) ...
Selecting previously unselected package libgs9-common.
Preparing to unpack .../05-libgs9-common 9.55.0~dfsg1-0ubuntu5.6_all.deb ...
Unpacking libgs9-common (9.55.0~dfsg1-Oubuntu5.6) ...
Selecting previously unselected package libidn12:amd64.
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Preparing to unpack .../06-libidn12_1.38-4ubuntu1_amd64.deb ...
Unpacking libidn12:amd64 (1.38-4ubuntu1) ...
Selecting previously unselected package libijs-0.35:amd64.
Preparing to unpack .../07-libijs-0.35_0.35-15build2_amd64.deb ...
Unpacking libijs-0.35:amd64 (0.35-15build2) ...
Selecting previously unselected package libjbig2dec0:amd64.
Preparing to unpack .../08-libjbig2dec0 0.19-3build2 amd64.deb ...
Unpacking libjbig2dec0:amd64 (0.19-3build2) ...
Selecting previously unselected package libgs9:amd64.
Preparing to unpack .../09-libgs9_9.55.0~dfsg1-Oubuntu5.6_amd64.deb ...
Unpacking libgs9:amd64 (9.55.0~dfsg1-Oubuntu5.6) ...
Selecting previously unselected package libkpathsea6:amd64.
Preparing to unpack .../10-libkpathsea6_2021.20210626.59705-1ubuntu0.2_amd64.deb
Unpacking libkpathsea6:amd64 (2021.20210626.59705-1ubuntu0.2) ...
Selecting previously unselected package libwoff1:amd64.
Preparing to unpack .../11-libwoff1_1.0.2-1build4_amd64.deb ...
Unpacking libwoff1:amd64 (1.0.2-1build4) ...
Selecting previously unselected package dvisvgm.
Preparing to unpack .../12-dvisvgm 2.13.1-1 amd64.deb ...
Unpacking dvisvgm (2.13.1-1) ...
Selecting previously unselected package fonts-lmodern.
Preparing to unpack .../13-fonts-lmodern_2.004.5-6.1_all.deb ...
Unpacking fonts-lmodern (2.004.5-6.1) ...
Selecting previously unselected package fonts-noto-mono.
Preparing to unpack .../14-fonts-noto-mono_20201225-1build1_all.deb ...
Unpacking fonts-noto-mono (20201225-1build1) ...
Selecting previously unselected package fonts-texgyre.
Preparing to unpack .../15-fonts-texgyre_20180621-3.1_all.deb ...
Unpacking fonts-texgyre (20180621-3.1) ...
Selecting previously unselected package libapache-pom-java.
Preparing to unpack .../16-libapache-pom-java_18-1_all.deb ...
Unpacking libapache-pom-java (18-1) ...
Selecting previously unselected package libcommons-parent-java.
Preparing to unpack .../17-libcommons-parent-java 43-1 all.deb ...
Unpacking libcommons-parent-java (43-1) ...
Selecting previously unselected package libcommons-logging-java.
Preparing to unpack .../18-libcommons-logging-java_1.2-2_all.deb ...
Unpacking libcommons-logging-java (1.2-2) ...
Selecting previously unselected package libfontenc1:amd64.
Preparing to unpack .../19-libfontenc1_1%3a1.1.4-1build3_amd64.deb ...
Unpacking libfontenc1:amd64 (1:1.1.4-1build3) ...
Selecting previously unselected package libptexenc1:amd64.
Preparing to unpack .../20-libptexenc1 2021.20210626.59705-1ubuntu0.2 amd64.deb
Unpacking libptexenc1:amd64 (2021.20210626.59705-1ubuntu0.2) ...
Selecting previously unselected package rubygems-integration.
Preparing to unpack .../21-rubygems-integration_1.18_all.deb ...
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Unpacking rubygems-integration (1.18) ...
Selecting previously unselected package ruby3.0.
Preparing to unpack .../22-ruby3.0_3.0.2-7ubuntu2.4_amd64.deb ...
Unpacking ruby3.0 (3.0.2-7ubuntu2.4) ...
Selecting previously unselected package ruby-rubygems.
Preparing to unpack .../23-ruby-rubygems_3.3.5-2_all.deb ...
Unpacking ruby-rubygems (3.3.5-2) ...
Selecting previously unselected package ruby.
Preparing to unpack .../24-ruby 1%3a3.0~exp1 amd64.deb ...
Unpacking ruby (1:3.0~exp1) ...
Selecting previously unselected package rake.
Preparing to unpack .../25-rake_13.0.6-2_all.deb ...
Unpacking rake (13.0.6-2) ...
Selecting previously unselected package ruby-net-telnet.
Preparing to unpack .../26-ruby-net-telnet_0.1.1-2_all.deb ...
Unpacking ruby-net-telnet (0.1.1-2) ...
Selecting previously unselected package ruby-webrick.
Preparing to unpack .../27-ruby-webrick_1.7.0-3_all.deb ...
Unpacking ruby-webrick (1.7.0-3) ...
Selecting previously unselected package ruby-xmlrpc.
Preparing to unpack .../28-ruby-xmlrpc_0.3.2-1ubuntu0.1_all.deb ...
Unpacking ruby-xmlrpc (0.3.2-1ubuntu0.1) ...
Selecting previously unselected package libruby3.0:amd64.
Preparing to unpack .../29-libruby3.0_3.0.2-7ubuntu2.4_amd64.deb ...
Unpacking libruby3.0:amd64 (3.0.2-7ubuntu2.4) ...
Selecting previously unselected package libsynctex2:amd64.
Preparing to unpack .../30-libsynctex2 2021.20210626.59705-1ubuntu0.2 amd64.deb
Unpacking libsynctex2:amd64 (2021.20210626.59705-1ubuntu0.2) ...
Selecting previously unselected package libteckit0:amd64.
Preparing to unpack .../31-libteckit0_2.5.11+ds1-1_amd64.deb ...
Unpacking libteckit0:amd64 (2.5.11+ds1-1) ...
Selecting previously unselected package libtexlua53:amd64.
Preparing to unpack .../32-libtexlua53_2021.20210626.59705-1ubuntu0.2_amd64.deb
Unpacking libtexlua53:amd64 (2021.20210626.59705-1ubuntu0.2) ...
Selecting previously unselected package libtexluajit2:amd64.
Preparing to unpack
.../33-libtexluajit2_2021.20210626.59705-1ubuntu0.2_amd64.deb ...
Unpacking libtexluajit2:amd64 (2021.20210626.59705-1ubuntu0.2) ...
Selecting previously unselected package libzzip-0-13:amd64.
Preparing to unpack .../34-libzzip-0-13_0.13.72+dfsg.1-1.1_amd64.deb ...
Unpacking libzzip-0-13:amd64 (0.13.72+dfsg.1-1.1) ...
Selecting previously unselected package xfonts-encodings.
Preparing to unpack .../35-xfonts-encodings 1%3a1.0.5-0ubuntu2 all.deb ...
Unpacking xfonts-encodings (1:1.0.5-Oubuntu2) ...
Selecting previously unselected package xfonts-utils.
Preparing to unpack .../36-xfonts-utils_1%3a7.7+6build2_amd64.deb ...
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Unpacking xfonts-utils (1:7.7+6build2) ...
Selecting previously unselected package lmodern.
Preparing to unpack .../37-lmodern_2.004.5-6.1_all.deb ...
Unpacking lmodern (2.004.5-6.1) ...
Selecting previously unselected package preview-latex-style.
Preparing to unpack .../38-preview-latex-style 12.2-1ubuntu1 all.deb ...
Unpacking preview-latex-style (12.2-1ubuntu1) ...
Selecting previously unselected package tlutils.
Preparing to unpack .../39-t1utils 1.41-4build2 amd64.deb ...
Unpacking tlutils (1.41-4build2) ...
Selecting previously unselected package teckit.
Preparing to unpack .../40-teckit_2.5.11+ds1-1_amd64.deb ...
Unpacking teckit (2.5.11+ds1-1) ...
Selecting previously unselected package tex-gyre.
Preparing to unpack .../41-tex-gyre_20180621-3.1_all.deb ...
Unpacking tex-gyre (20180621-3.1) ...
Selecting previously unselected package texlive-binaries.
Preparing to unpack .../42-texlive-
binaries_2021.20210626.59705-1ubuntu0.2_amd64.deb ...
Unpacking texlive-binaries (2021.20210626.59705-1ubuntu0.2) ...
Selecting previously unselected package texlive-base.
Preparing to unpack .../43-texlive-base 2021.20220204-1 all.deb ...
Unpacking texlive-base (2021.20220204-1) ...
Selecting previously unselected package texlive-fonts-recommended.
Preparing to unpack .../44-texlive-fonts-recommended_2021.20220204-1_all.deb ...
Unpacking texlive-fonts-recommended (2021.20220204-1) ...
Selecting previously unselected package texlive-latex-base.
Preparing to unpack .../45-texlive-latex-base 2021.20220204-1_all.deb ...
Unpacking texlive-latex-base (2021.20220204-1) ...
Selecting previously unselected package libfontbox-java.
Preparing to unpack .../46-libfontbox-java_1%3a1.8.16-2_all.deb ...
Unpacking libfontbox-java (1:1.8.16-2) ...
Selecting previously unselected package libpdfbox-java.
Preparing to unpack .../47-libpdfbox-java_1%3a1.8.16-2_all.deb ...
Unpacking libpdfbox-java (1:1.8.16-2) ...
Selecting previously unselected package texlive-latex-recommended.
Preparing to unpack .../48-texlive-latex-recommended 2021.20220204-1 all.deb ...
Unpacking texlive-latex-recommended (2021.20220204-1) ...
Selecting previously unselected package texlive-pictures.
Preparing to unpack .../49-texlive-pictures_2021.20220204-1_all.deb ...
Unpacking texlive-pictures (2021.20220204-1) ...
Selecting previously unselected package texlive-latex-extra.
Preparing to unpack .../50-texlive-latex-extra_2021.20220204-1_all.deb ...
Unpacking texlive-latex-extra (2021.20220204-1) ...
Selecting previously unselected package texlive-plain-generic.
Preparing to unpack .../51-texlive-plain-generic_2021.20220204-1_all.deb ...
Unpacking texlive-plain-generic (2021.20220204-1) ...
Selecting previously unselected package tipa.
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Preparing to unpack .../52-tipa_2%3a1.3-21_all.deb ...
Unpacking tipa (2:1.3-21) ...
Selecting previously unselected package texlive-xetex.
Preparing to unpack .../53-texlive-xetex_2021.20220204-1_all.deb ...
Unpacking texlive-xetex (2021.20220204-1) ...
Setting up fonts-lato (2.0-2.1) ...
Setting up fonts-noto-mono (20201225-1build1) ...
Setting up libwoff1:amd64 (1.0.2-1build4) ...
Setting up libtexlua53:amd64 (2021.20210626.59705-1ubuntu0.2) ...
Setting up libijs-0.35:amd64 (0.35-15build2) ...
Setting up libtexluajit2:amd64 (2021.20210626.59705-1ubuntu0.2) ...
Setting up libfontbox-java (1:1.8.16-2) ...
Setting up rubygems-integration (1.18) ...
Setting up libzzip-0-13:amd64 (0.13.72+dfsg.1-1.1) ...
Setting up fonts-urw-base35 (20200910-1) ...
Setting up poppler-data (0.4.11-1) ...
Setting up tex-common (6.17) ...
debconf: unable to initialize frontend: Dialog
debconf: (No usable dialog-like program is installed, so the dialog based
frontend cannot be used. at /usr/share/perl5/Debconf/FrontEnd/Dialog.pm line
78.)
debconf: falling back to frontend: Readline
update-language: texlive-base not installed and configured, doing nothing!
Setting up libfontenc1:amd64 (1:1.1.4-1build3) ...
Setting up libjbig2dec0:amd64 (0.19-3build2) ...
Setting up libteckit0:amd64 (2.5.11+ds1-1) ...
Setting up libapache-pom-java (18-1) ...
Setting up ruby-net-telnet (0.1.1-2) ...
Setting up xfonts-encodings (1:1.0.5-Oubuntu2) ...
Setting up t1utils (1.41-4build2) ...
Setting up libidn12:amd64 (1.38-4ubuntu1) ...
Setting up fonts-texgyre (20180621-3.1) ...
Setting up libkpathsea6:amd64 (2021.20210626.59705-1ubuntu0.2) ...
Setting up ruby-webrick (1.7.0-3) ...
Setting up fonts-lmodern (2.004.5-6.1) ...
Setting up fonts-droid-fallback (1:6.0.1r16-1.1build1) ...
Setting up ruby-xmlrpc (0.3.2-1ubuntu0.1) ...
Setting up libsynctex2:amd64 (2021.20210626.59705-1ubuntu0.2) ...
Setting up libgs9-common (9.55.0~dfsg1-Oubuntu5.6) ...
Setting up teckit (2.5.11+ds1-1) ...
Setting up libpdfbox-java (1:1.8.16-2) ...
Setting up libgs9:amd64 (9.55.0~dfsg1-Oubuntu5.6) ...
Setting up preview-latex-style (12.2-1ubuntu1) ...
Setting up libcommons-parent-java (43-1) ...
Setting up dvisvgm (2.13.1-1) ...
Setting up libcommons-logging-java (1.2-2) ...
Setting up xfonts-utils (1:7.7+6build2) ...
Setting up libptexenc1:amd64 (2021.20210626.59705-1ubuntu0.2) ...
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Setting up texlive-binaries (2021.20210626.59705-1ubuntu0.2) ...
update-alternatives: using /usr/bin/xdvi-xaw to provide /usr/bin/xdvi.bin
(xdvi.bin) in auto mode
update-alternatives: using /usr/bin/bibtex.original to provide /usr/bin/bibtex
(bibtex) in auto mode
Setting up lmodern (2.004.5-6.1) ...
Setting up texlive-base (2021.20220204-1) ...
/usr/bin/ucfr
/usr/bin/ucfr
/usr/bin/ucfr
/usr/bin/ucfr
mktexlsr: Updating /var/lib/texmf/ls-R-TEXLIVEDIST...
mktexlsr: Updating /var/lib/texmf/ls-R-TEXMFMAIN...
mktexlsr: Updating /var/lib/texmf/ls-R...
mktexlsr: Done.
tl-paper: setting paper size for dvips to a4:
/var/lib/texmf/dvips/config/config-paper.ps
tl-paper: setting paper size for dvipdfmx to a4:
/var/lib/texmf/dvipdfmx/dvipdfmx-paper.cfg
tl-paper: setting paper size for xdvi to a4: /var/lib/texmf/xdvi/XDvi-paper
tl-paper: setting paper size for pdftex to a4: /var/lib/texmf/tex/generic/tex-
ini-files/pdftexconfig.tex
debconf: unable to initialize frontend: Dialog
debconf: (No usable dialog-like program is installed, so the dialog based
frontend cannot be used. at /usr/share/perl5/Debconf/FrontEnd/Dialog.pm line
78.)
debconf: falling back to frontend: Readline
Setting up tex-gyre (20180621-3.1) ...
Setting up texlive-plain-generic (2021.20220204-1) ...
Setting up texlive-latex-base (2021.20220204-1) ...
Setting up texlive-latex-recommended (2021.20220204-1) ...
Setting up texlive-pictures (2021.20220204-1) ...
Setting up texlive-fonts-recommended (2021.20220204-1) ...
Setting up tipa (2:1.3-21) ...
Setting up texlive-latex-extra (2021.20220204-1) ...
Setting up texlive-xetex (2021.20220204-1) ...
Setting up rake (13.0.6-2) ...
Setting up libruby3.0:amd64 (3.0.2-7ubuntu2.4) ...
Setting up ruby3.0 (3.0.2-7ubuntu2.4) ...
Setting up ruby (1:3.0~exp1) ...
Setting up ruby-rubygems (3.3.5-2) ...
Processing triggers for man-db (2.10.2-1) ...
Processing triggers for fontconfig (2.13.1-4.2ubuntu5) ...
Processing triggers for libc-bin (2.35-Oubuntu3.4) ...
/sbin/ldconfig.real: /usr/local/lib/libtbb.so.12 is not a symbolic link
/sbin/ldconfig.real: /usr/local/lib/libtbbbind_2_5.so.3 is not a symbolic link
```

/sbin/ldconfig.real: /usr/local/lib/libtbbbind.so.3 is not a symbolic link

/sbin/ldconfig.real: /usr/local/lib/libtbbmalloc_proxy.so.2 is not a symbolic

link

/sbin/ldconfig.real: /usr/local/lib/libtbbmalloc.so.2 is not a symbolic link

/sbin/ldconfig.real: /usr/local/lib/libtbbbind_2_0.so.3 is not a symbolic link

Processing triggers for tex-common (6.17) ...

debconf: unable to initialize frontend: Dialog

debconf: (No usable dialog-like program is installed, so the dialog based frontend cannot be used. at /usr/share/perl5/Debconf/FrontEnd/Dialog.pm line 78.)

debconf: falling back to frontend: Readline

Running updmap-sys. This may take some time... done.

Running mktexlsr /var/lib/texmf ... done.

Building format(s) --all.

This may take some time... done.