The Wall Street Journal collected data about a student's major, type of college attended and region of college and their career earnings at starting and mid career of their careers. They also went ahead and collected data about the salary collected at various percentiles during mid career. In this project, I conducted an EDA to determine how each of the above mentioned factors contribute to a student's prospective career earning. College applicants consider salary as an important factor before deciding which college path to follow. Objective Identify how undergraduate major, college type and college region impacts an individual's career earning potential and help individuals make an informed decision. Collecting the Data The data required for this project has been provided from a Kaggle notebook. The data is collected and owned by the Wall Street Journal, and since it is a reputable organization, the data can be considered reliable and able to assist us in this project. The data only contains information about colleges and universities located within the United States of America making it less ideal for college applicants who plan to pursue careers in other countries. Data also does not contain college fees required for each course leaving us unable to determine which course, college and region offer the highest value for money. Processing the Data For processing, analyzing and visualizing the data we need to import some key Python libraries In [1]: import pandas as pd import numpy as np from matplotlib import pyplot as plt plt.style.use('fivethirtyeight') First we need to import the data degrees_df = pd.read_csv("/Users/arun/Documents/Portfolio Projects/College Data/degrees.csv") In [2]: In [3]: degrees_df **Starting Median** Mid-Career Median Percent change from Starting to Mid-Career 10th Mid-Career 25th Mid-Career 75th Mid-Career 90th Out[3]: **Undergraduate Major** Salary Salary Mid-Career Salary **Percentile Salary Percentile Salary** Percentile Salary **Percentile Salary** \$152,000.00 0 Accounting \$46,000.00 \$77,100.00 67.6 \$42,200.00 \$56,100.00 \$108,000.00 \$57,700.00 \$101,000.00 75.0 \$64,300.00 \$82,100.00 \$127,000.00 \$161,000.00 Aerospace Engineering 2 \$96,300.00 \$150,000.00 \$42,600.00 \$71,900.00 \$36,300.00 \$52,100.00 Agriculture 68.8 \$36,800.00 \$61,500.00 67.1 \$33,800.00 \$45,500.00 \$89,300.00 \$138,000.00 Anthropology \$41,600.00 \$76,800.00 84.6 \$50,600.00 \$62,200.00 \$97,000.00 \$136,000.00 4 Architecture Art History \$35,800.00 \$64,900.00 81.3 \$28,800.00 \$42,200.00 \$87,400.00 \$125,000.00 6 \$38,800.00 \$64,800.00 67.0 \$36,900.00 \$47,400.00 \$94,500.00 \$135,000.00 Biology \$72,100.00 \$51,500.00 **Business Management** \$43,000.00 67.7 \$38,800.00 \$102,000.00 \$147,000.00 8 \$63,200.00 \$107,000.00 69.3 \$71,900.00 \$87,300.00 \$143,000.00 \$194,000.00 Chemical Engineering \$42,600.00 \$79,900.00 87.6 \$45,300.00 \$60,700.00 \$108,000.00 \$148,000.00 Chemistry 10 Civil Engineering \$53,900.00 \$90,500.00 67.9 \$63,400.00 \$75,100.00 \$115,000.00 \$148,000.00 11 Communications \$38,100.00 \$70,000.00 83.7 \$37,500.00 \$49,700.00 \$98,800.00 \$143,000.00 12 \$61,400.00 \$105,000.00 71.0 \$66,100.00 \$84,100.00 \$135,000.00 \$162,000.00 Computer Engineering \$56,000,00 \$154.000.00 13 Computer Science \$55,900.00 \$95,500.00 \$74,900.00 \$122,000.00 \$171,000.00 \$56,300.00 \$68,100.00 \$118,000.00 14 \$53,700.00 \$88,900.00 65.5 Construction Criminal Justice \$56,300.00 \$80,700.00 15 \$35,000.00 60.9 \$32,200.00 \$41,600.00 \$107,000.00 16 \$35,900.00 \$56,900.00 58.5 \$36,700.00 \$41,300.00 \$79,100.00 \$153,000.00 Drama \$50,100.00 \$50,600.00 \$210,000.00 17 **Economics** \$98,600.00 96.8 \$70,600.00 \$145,000.00 \$73,400.00 18 \$34,900.00 \$52,000.00 49.0 \$29,300.00 \$37,900.00 \$102,000.00 Education **Electrical Engineering** \$83,800.00 19 \$60,900.00 \$103,000.00 69.1 \$69,300.00 \$130,000.00 \$168,000.00 20 \$38,000.00 \$64,700.00 70.3 \$33,400.00 \$44,800.00 \$93,200.00 \$133,000.00 English 21 \$37,900.00 \$68,500.00 80.7 \$33,900.00 \$45,500.00 \$100,000.00 \$136,000.00 22 \$47,900.00 \$88,300.00 84.3 \$47,200.00 \$62,100.00 \$128,000.00 \$195,000.00 Finance \$49,300.00 \$78,200.00 23 \$39,100.00 \$62,600.00 60.1 \$41,000.00 \$111,000.00 Forestry 59.0 \$40,000.00 \$132,000.00 24 \$41,200.00 \$65,500.00 \$50,000.00 \$90,800.00 Geography \$79,500.00 \$45,000.00 \$59,600.00 25 Geology \$43,500.00 82.8 \$101,000.00 \$156,000.00 26 Graphic Design \$35,700.00 \$59,800.00 67.5 \$36,000.00 \$45,500.00 \$80,800.00 \$112,000.00 27 Health Care Administration \$38,800.00 \$60,600.00 56.2 \$34,600.00 \$45,600.00 \$78,800.00 \$101,000.00 28 81.1 \$37,000.00 \$49,200.00 \$103,000.00 \$149,000.00 \$39,200.00 \$71,000.00 History 29 Hospitality & Tourism \$37,800.00 \$57,500.00 52.1 \$35,500.00 \$43,600.00 \$81,900.00 \$124,000.00 30 \$57,700.00 64.1 \$57,100.00 \$72,300.00 \$132,000.00 \$173,000.00 Industrial Engineering \$94,700.00 31 \$49,100.00 \$74,800.00 52.3 \$44,500.00 \$56,700.00 \$96,700.00 \$129,000.00 Information Technology (IT) 32 47.4 \$35,700.00 \$42,600.00 \$72,500.00 \$107,000.00 \$36,100.00 \$53,200.00 Interior Design 33 International Relations \$40,900.00 \$80,900.00 97.8 \$38,200.00 \$56,000.00 \$111,000.00 \$157,000.00 34 \$38,400.00 \$48,300.00 \$97,700.00 \$145,000.00 \$35,600.00 \$66,700.00 87.4 Journalism Management Information 35 \$49,200.00 \$82,300.00 67.3 \$45,300.00 \$60,500.00 \$108,000.00 \$146,000.00 Systems (MIS) 36 Marketing \$40,800.00 \$79,600.00 95.1 \$42,100.00 \$55,600.00 \$119,000.00 \$175,000.00 37 \$45,400.00 \$92,400.00 103.5 \$45,200.00 \$64,200.00 \$128,000.00 \$183,000.00 Math 38 Mechanical Engineering \$57,900.00 \$93,600.00 61.7 \$63,700.00 \$76,200.00 \$120,000.00 \$163,000.00 \$26,700.00 \$40,200.00 \$88,000.00 \$134,000.00 39 \$35,900.00 \$55,000.00 53.2 Music 40 \$54,200.00 \$67,000.00 23.6 \$47,600.00 \$56,400.00 \$80,900.00 \$98,300.00 Nursing 38.6 \$33,900.00 \$44,500.00 \$70,500.00 \$99,200.00 41 Nutrition \$39,900.00 \$55,300.00 42 \$39,900.00 \$81,200.00 103.5 \$35,500.00 \$52,800.00 \$127,000.00 \$168,000.00 Philosophy \$124,000.00 43 \$74,300.00 \$91,700.00 23.4 \$66,400.00 \$75,200.00 \$108,000.00 Physician Assistant 44 \$50,300.00 \$97,300.00 93.4 \$56,000.00 \$74,200.00 \$132,000.00 \$178,000.00 Physics 91.7 45 \$40,800.00 \$78,200.00 \$41,200.00 \$55,300.00 \$114,000.00 \$168,000.00 Political Science \$60,400.00 46 Psychology \$35,900.00 68.2 \$31,600.00 \$42,100.00 \$87,500.00 \$127,000.00 47 \$34,100.00 \$52,000.00 52.5 \$29,700.00 \$36,500.00 \$70,900.00 \$96,400.00 Religion \$58,200.00 \$30,700.00 \$40,400.00 \$81,200.00 48 \$36,500.00 59.5 \$118,000.00 Sociology 56.2 \$31,000.00 \$40,000.00 \$76,800.00 \$96,400.00 \$34,000.00 \$53,100.00 Spanish Similarly, we need to import the remaining 2 datasets college_type_df = pd.read_csv("/Users/arun/Documents/Portfolio Projects/College Data/college_type.csv") region_df = pd.read_csv("/Users/arun/Documents/Portfolio Projects/College Data/region.csv") In order to perform analysis on the data, the salaries have to be converted into 'float' datatype degrees_df[degrees_df.columns[1:3]] = degrees_df[degrees_df.columns[1:3]].replace('[\$,]', '', regex=True).astype(float) degrees_df[degrees_df.columns[4:]] = degrees_df[degrees_df.columns[4:]].replace('[\$,]', '', regex=True).astype(float) To see if the data type conversion was successful, we check using a preview of the data In [8]: degrees_df.head() **Starting Median** Undergraduate Mid-Career Median Percent change from Starting to Mid-Mid-Career 10th Percentile Mid-Career 25th Percentile Mid-Career 75th Percentile Mid-Career 90th Percentile Out[8]: **Career Salary** Major Salary Salary Salary Salary Salary Salary 0 Accounting 46000.0 77100.0 67.6 42200.0 56100.0 108000.0 152000.0 Aerospace 57700.0 101000.0 75.0 64300.0 82100.0 127000.0 161000.0 Engineering 2 Agriculture 42600.0 71900.0 68.8 36300.0 52100.0 96300.0 150000.0 3 36800.0 61500.0 67.1 33800.0 45500.0 89300.0 138000.0 Anthropology 41600.0 76800.0 50600.0 62200.0 4 Architecture 84.6 97000.0 136000.0 college_type_df[college_type_df.columns[2:]] = college_type_df[college_type_df.columns[2:]].replace('[\$,]','', regex=True).astype(float) In [9]: college_type_df.head() In [10]: **Starting Median** Mid-Career 90th Percentile **School** Mid-Career Median **Mid-Career 10th Percentile Mid-Career 25th Percentile Mid-Career 75th Percentile** Out[10]: **School Name** Type Salary Salary Salary Salary Salary Salary Massachusetts Institute of Technology Engineering 76800.0 99200.0 220000.0 72200.0 126000.0 168000.0 (MIT) California Institute of Technology (CIT) Engineering 75500.0 123000.0 NaN 104000.0 161000.0 NaN 2 Harvey Mudd College 71800.0 122000.0 NaN 96000.0 180000.0 NaN Engineering Polytechnic University of New York, 3 Engineering 62400.0 114000.0 66800.0 94300.0 143000.0 190000.0 Brooklyn 4 62200.0 114000.0 80200.0 142000.0 Cooper Union Engineering NaN NaN region_df[region_df.columns[2:]] = region_df[region_df.columns[2:]].replace('[\$,]','',regex=True).astype(float) region_df.head() In [12] Mid-Career 10th Percentile Mid-Career 75th Percentile **Starting Median** Mid-Career Median Mid-Career 25th Percentile Mid-Career 90th Percentile Out[12]: School Name Region Salary Salary Salary Salary Salary Salary 0 Stanford University California 70400.0 129000.0 68400.0 93100.0 184000.0 257000.0 California Institute of Technology California 75500.0 123000.0 104000.0 161000.0 NaN NaN (CIT) 2 Harvey Mudd College California 71800.0 122000.0 NaN 96000.0 180000.0 NaN 3 University of California, Berkeley California 59900.0 112000.0 59500.0 81000.0 149000.0 201000.0 4 51900.0 105000.0 54800.0 157000.0 Occidental College California NaN NaN The College Type dataset and Region dataset both contain some missing values. These values must be removed in order to do analysis. If possible, more research would be done to obtain those values but since that is not feasible, the rows containing missing values will be deleted. college_type_df.dropna(axis=0, how='any', thresh=None, subset=None, inplace=False) In [13]: Mid-Career 10th Percentile Mid-Career 25th Percentile **Mid-Career 75th Percentile** Mid-Career 90th Percentile Out[13]: School **Starting Median** Mid-Career Median **School Name** Type Salary Salary Salary Salary Salary Salary Massachusetts Institute of Technology 0 72200.0 126000.0 76800.0 99200.0 168000.0 220000.0 Engineering (MIT) Polytechnic University of New York, 3 62400.0 114000.0 66800.0 94300.0 143000.0 190000.0 Engineering Brooklyn Worcester Polytechnic Institute (WPI) 61000.0 114000.0 0.00008 91200.0 137000.0 180000.0 Engineering 61800.0 Carnegie Mellon University (CMU) 111000.0 63300.0 80100.0 150000.0 209000.0 Engineering Rensselaer Polytechnic Institute (RPI) 61100.0 110000.0 71600.0 85500.0 140000.0 182000.0 Engineering Austin Peay State University 265 40400.0 58200.0 25600.0 46000.0 84600.0 117000.0 Pittsburg State University State 266 Southern Utah University 41900.0 56500.0 30700.0 39700.0 78400.0 116000.0 State 37900.0 50600.0 22600.0 31800.0 78500.0 98900.0 267 Montana State University - Billings State 268 Black Hills State University State 35300.0 43900.0 27000.0 32200.0 60900.0 87600.0 231 rows × 8 columns region_df.dropna(axis=0, how='any', thresh=None, subset=None, inplace=False) In [14]: **Starting Median Mid-Career Median** Mid-Career 10th Percentile Mid-Career 25th Percentile Mid-Career 75th Percentile Mid-Career 90th Percentile Out[14]: **School Name** Region Salary Salary Salary Salary Salary Salary 0 93100.0 184000.0 257000.0 Stanford University California 70400.0 129000.0 68400.0 3 University of California, Berkeley California 59900.0 112000.0 59500.0 81000.0 149000.0 201000.0 5 Cal Poly San Luis Obispo 57200.0 101000.0 55000.0 74700.0 133000.0 178000.0 California University of California at Los Angeles California 52600.0 101000.0 51300.0 72500.0 139000.0 193000.0 University of California, San Diego 7 California 51100.0 101000.0 51700.0 75400.0 131000.0 177000.0 (UCSD) State University of New York (SUNY) at 315 Northeastern 38000.0 70300.0 35100.0 51200.0 100000.0 179000.0 Potsdam 36900.0 69700.0 44000.0 57000.0 92000.0 128000.0 316 Niagara University Northeastern State University of New York (SUNY) at 317 Northeastern 37800.0 66200.0 32800.0 44200.0 93300.0 181000.0 39400.0 63600.0 40400.0 47900.0 85700.0 117000.0 318 University of Southern Maine Northeastern 35600.0 47300.0 99000.0 134000.0 319 Mercy College Northeastern 43700.0 62600.0 273 rows × 8 columns Analyzing the Data Analyzing Salaries by Undergraduate Major First, we are gonna determine which undergraduate degree offers the highest starting median salary high_major_start_sal = degrees_df.loc[degrees_df["Starting Median Salary"].idxmax()] In [15]: high_major_start_sal In [16]: Undergraduate Major Physician Assistant Out[16]: Starting Median Salary 74300.0 91700.0 Mid-Career Median Salary Percent change from Starting to Mid-Career Salary 23.4 Mid-Career 10th Percentile Salary 66400.0 Mid-Career 25th Percentile Salary 75200.0 Mid-Career 75th Percentile Salary 108000.0 Mid-Career 90th Percentile Salary 124000.0 Name: 43, dtype: object degrees_sorted_start = degrees_df.sort_values('Starting Median Salary', ascending=False).head(10) In [17]: start_sal = degrees_sorted_start['Starting Median Salary'].head(10) start_sal = start_sal.iloc[::-1] In [19]: undergrad_major_start = degrees_sorted_start['Undergraduate Major'].head(10) undergrad_major_start = undergrad_major_start.iloc[::-1] In [22]: plt.barh(undergrad_major_start, start_sal) plt.title('Top 10 Jobs by Starting Median Salary') plt.xlabel('Salary in USD') plt.show() Top 10 Jobs by Starting Median Salary Physician Assistant Chemical Engineering Computer Engineering Electrical Engineering Mechanical Engineering Aerospace Engineering Industrial Engineering Computer Science Nursing Civil Engineering 0 10000 20000 30000 40000 50000 60000 70000 Salary in USD Next, we will find the highest paying career at mid-career high_major_mid_sal = degrees_df.loc[degrees_df["Mid-Career Median Salary"].idxmax()] In [23]:

Analysis of Career Salaries by Major, College Type and Region

Introduction

high_major_mid_sal

Undergraduate Major

Starting Median Salary

Name: 8, dtype: object

107000.0

105000.0

103000.0

101000.0

98600.0

97300.0

95500.0

94700.0

93600.0

92400.0

mid_sal = mid_sal.iloc[::-1]

plt.xlabel('Salary in USD')

Chemical Engineering Computer Engineering Electrical Engineering Aerospace Engineering

plt.show()

mid_sal

12 19

1

17

44

13

30

38

37

Mid-Career Median Salary

Mid-Career 10th Percentile Salary

Mid-Career 25th Percentile Salary

Mid-Career 75th Percentile Salary

Mid-Career 90th Percentile Salary

Percent change from Starting to Mid-Career Salary

Name: Mid-Career Median Salary, dtype: float64

plt.barh(undergrad_major_mid, mid_sal)

Economics Physics

0

Analyzing Salaries by College Type

59057.894737

60475.000000

45746.808511

45715.000000

44126.285714 Name: Starting Median Salary, dtype: float64

plt.bar(college_type, college_type_start_plot) plt.title("Starting Median Salary by Collge Type")

Next, we will find the highest mid-career salary by college type

103842.105263

120125.000000

89378.723404

84685.000000 78567.428571

Name: Mid-Career Median Salary, dtype: float64

plt.bar(college_type, college_type_mid_mean)

Analyzing Salaries by College Region

51032.142857

48496.000000 44521.518987

44414.285714

44225.352113

Name: Starting Median Salary, dtype: float64

plt.title("Starting Median Salary by Region")

Now to find the mean mid-career salaries by region

93132.142857 91352.000000

79505.063291

78200.000000

78180.281690 Name: Mid-Career Median Salary, dtype: float64

plt.title("Mid-Career Median by Region")

starting median salaries and mid-career median salaries.

plt.bar(regions, region_mid_mean)

plt.ylabel("Salary in USD") plt.xticks(rotation=45)

Starting Median Salary by Region

ADD GRAPH ABOUT START MED SAL BY REGION

Mid-Career Median by Region

region_mid_mean = region_df.groupby("Region")["Mid-Career Median Salary"].mean().sort_values().iloc[::-1]

plt.bar(regions, region_start_mean)

plt.ylabel("Salary in USD") plt.xticks(rotation=45)

regions = region_df['Region'].unique()

plt.ylabel("Salary in USD") plt.xticks(rotation=45)

plt.title("Mid-Career Median Salary by College Type")

college_type_mid_mean = college_type_mid_mean.sort_values().iloc[::-1]

Mid-Career Median Salary by College Type

regions = ['California', 'Northeastern', 'Southern', 'Western', 'Midwestern']

region_start_mean = (region_df.groupby("Region")["Starting Median Salary"].mean()).sort_values().iloc[::-1]

20000

college_type = college_type_df["School Type"].unique()

40000

college_type = ['Ivy Leauge', 'Engineering', 'Liberal Arts', 'Party', 'State']

college_type_start_plot = college_type_start_mean.sort_values().iloc[::-1]

Starting Median Salary by Collge Type

Computer Science Industrial Engineering Mechanical Engineering

college_type_start_mean

plt.ylabel("Salary in USD") plt.xticks(rotation=45)

School Type

Engineering

Liberal Arts

Ivy League

plt.show()

60000

50000

Salary in USD 40000 20000

10000

In [39]:

Out[39]:

In [40]:

0

college_type_mid_mean

School Type

Engineering

Ivy League

plt.show()

120000

100000

80000

60000

40000

20000

0

region_start_mean

Region

California

Midwestern

plt.show()

50000

40000

30000

20000

10000

0

region_mid_mean

Region

California

Midwestern

plt.show()

80000

60000

40000

20000

0

Conclusion

Salary in USD

Southern

Northeastern

Out[48]

Salary in USD

Southern

Western

Northeastern

Salary in USD

In [42]:

In [43]:

In [44]:

In [45]:

Out[45]

In [46]:

Party

State

Liberal Arts

Party

State

undergrad_major_mid = undergrad_major_mid.iloc[::-1]

plt.title('Top 10 Jobs by Mid-Career Median Salary')

mid_sal = degrees_sorted_mid['Mid-Career Median Salary'].head(10)

undergrad_major_mid = degrees_sorted_mid['Undergraduate Major'].head(10)

Top 10 Jobs by Mid-Career Median Salary

60000

college_type_start_mean = college_type_df.groupby("School Type")["Starting Median Salary"].mean()

college_type_mid_mean = college_type_df.groupby("School Type")["Mid-Career Median Salary"].mean()

Salary in USD

80000

100000

Now, we will find the which college type offers the highest starting median salary. In order to that, we will need to group all colleges of the similar type together and take the mean.

In order to get a clear image on salaries by region, we need to group colleges by the region they are present in and then find mean. First, we will find the mean starting median salary by college region

From the analysis of the data, it is evident Physician's Assistant offer the highest starting median salary. For individuals looking to earn more money immediately after graduation, it is recommended they pursue

an Undergraduate Major to become Physician Assistant. If they are looking long term, then Chemical Engineering offers the highest median mid-career salary. If an individual is using College Type as the deciding factor, Ivy League schools offer both the highest starting median salaries and mid-career median salaries. If the individual is looking for colleges by region, California regions offering the highest

degrees_sorted_mid = degrees_df.sort_values('Mid-Career Median Salary', ascending=False)

Out[24]:

Out[27]:

In [30]:

In [32]:

In [33]:

Out[35]:

Chemical Engineering

63200.0

69.3

107000.0

71900.0

87300.0

143000.0

194000.0