

# Final Project: Is College Worth It? (Tuition VS Salary)

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## I. Introduction

Is college worth it? Is college a good investment for your future? If it is, what kind of factors in college would have an impact on career performance?

On one hand, college could be worth it by leading to higher employment rates and higher career performance, in terms of various financial measurements, than people who do not go to college. On the other hand, college tuition is constantly rising and is the same for student loan debt.

In this project, four data sources are acquired from the US Department of Education, the Chronicle of Higher Education, the National Center for Education Statistics, and payscale.com. A final dataset in tidy version is created by conducting a significant amount of data cleansing and data wrangling techniques, so as to retrieve insightful information regarding the relationship between tuition or other factors in college and future career performance of college graduates.

### Github Link:

(<https://github.com/Junjie-Dylan-Yang/Data-Wrangling-Project>)

## II. ETL process: Data Import and Data Cleansing

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### Import first data: tuition\_cost

First data source, tuition\_cost, is from “College tuition, Diversity, and Pay” in rfordatascience/tidetuesday/2020-03-10, which is originally acquired from the US Department of Education and the Chronicle of Higher Education.

### Data Cleaning for tuition\_cost data

In the tuition\_cost data, relevant columns are selected (name of the school, state, state code, type of the school, length of the degree). Also, room and board fee and tuition are combined as total tuition and fee.

### Below is the snippet of tuition\_cost data

```
## # A tibble: 10 x 7
##   name    state state_code type  degree_length in_state_tuitio~ out_of_state_tu~
##   <chr>   <chr> <chr>      <chr> <chr>          <dbl>          <dbl>
## 1 Aanii~ Mont~ MT        Publ~ 2 Year          2380           2380
## 2 Abile~ Texas TX        Priv~ 4 Year          45200          45200
## 3 Abrah~ Geor~ GA        Publ~ 2 Year          12602          21024
## 4 Acade~ Minn~ MN        For ~ 2 Year          17661          17661
## 5 Acade~ Cali~ CA        For ~ 4 Year          44458          44458
```

##	6	Adams~ Colo~ CO	Publ~ 4 Year	18222	29238
##	7	Adelp~ New ~ NY	Priv~ 4 Year	54690	54690
##	8	Adiro~ New ~ NY	Publ~ 2 Year	17035	21595
##	9	Adria~ Mich~ MI	Priv~ 4 Year	48405	48405
##	10	Advan~ Virg~ VA	For ~ 2 Year	13680	13680

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## Import second data: student\_diversity

Second data source, student\_diversity by college/university, along with school type, degree length, state, in-state vs out-of-state is from the Chronicle of Higher Education.

## Data Cleaning for student\_diversity data

In the student\_diversity data, the main data cleansing task is to modify name of institution to match the “name” column and “state” column in the tuition\_cost data, in order to combine dataset. Several data wrangling steps were applied. First is to change the column name “INSTITUTION” to “name”. After that, convert any abbreviation of University from “U.” to “University”. From the first glance, the name of state is located at the very end of the name of institution. The next step is to extract state from school name with the help of state.name which contains the list of all the state name and column “state” is created. Last but not least, state name inside the name of institution needed to remove. Using str\_count to count the letters within state in each observation and str\_sub help to keep the name of school only in the “name” column. Str\_trim and str\_squish are used to remove unnecessary spaces in “name”.

## Below is the snippet of student\_diversity\_cleaned data

```
## # A tibble: 10 x 11
##   name ENROLLMENT WOMEN `AMERICAN INDIA~ ASIAN BLACK HISPANIC
##   <chr>      <dbl>  <dbl>      <dbl> <dbl> <dbl> <dbl>
## 1 Univ~    195059 134722      876  1959 31455 13984
## 2 Ivy ~     91179  53476      357  1369 12370  5533
## 3 Libe~     81459  48329      447   856 14751  1186
## 4 Lone~     69395  41268      168  4198 12094 23751
## 5 Miam~     66046  38323       47   655 10722 44870
## 6 Gran~     62304  46647      586  2446 13856  8933
## 7 Texa~     61642  29277      173  3545  1879 11256
## 8 Univ~     60767  33482      120  3343  6400 13108
## 9 Ohio~     58322  28658       76  3339  3108  2049
## 10 Hous~     58276  34007      116  5391 18520 18411
## # ... with 4 more variables: `NATIVE HAWAIIAN / PACIFIC ISLANDER` <dbl>,
## #   WHITE <dbl>, `TOTAL MINORITY` <dbl>, state <chr>
```

## Combine tuition\_cost and student\_diversity data based on “name” and “state”

So far, student\_diversity and tuition\_cost are modified to share two common column, “name” – name of the school and “state” – the state that the school is located. Thus, student\_diversity and tuition\_cost datasets are merged for late development. There are a few schools appears in the tuition\_cost dataset but not in the student\_diversity and “NA” value appear. It is reasonable and schools with “NA” value are removed from the combined dataset. The combined dataset is arranged by state and the name of the school.

Below is the snippet of the combined dataset, `tuition_with_diversity`

```
## # A tibble: 10 x 16
##   name state state_code type degree_length in_state_tuitio~ out_of_state_tu~
##   <chr> <chr> <chr>      <chr> <chr>                <dbl>          <dbl>
## 1 Alab~ Alab~ AL        Publ~ 2 Year              4440           8880
## 2 Alab~ Alab~ AL        Publ~ 4 Year             16490          24818
## 3 Amri~ Alab~ AL        Priv~ 4 Year              6900           6900
## 4 Athe~ Alab~ AL        Publ~ 4 Year              6810          12870
## 5 Aubu~ Alab~ AL        Publ~ 4 Year             24608          43856
## 6 Aubu~ Alab~ AL        Publ~ 4 Year             17268          29028
## 7 Bevi~ Alab~ AL        Publ~ 2 Year              6070           9940
## 8 Birm~ Alab~ AL        Priv~ 4 Year             30000          30000
## 9 Bish~ Alab~ AL        Publ~ 2 Year              4740           8610
## 10 Calh~ Alab~ AL        Publ~ 2 Year              4840           8690
## # ... with 9 more variables: ENROLLMENT <dbl>, WOMEN <dbl>, `AMERICAN INDIAN /
## #   ALASKA NATIVE` <dbl>, ASIAN <dbl>, BLACK <dbl>, HISPANIC <dbl>, `NATIVE
## #   HAWAIIAN / PACIFIC ISLANDER` <dbl>, WHITE <dbl>, `TOTAL MINORITY` <dbl>
```

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### Import third data: `Best_School`

Third data source, `best_school` is html data, acquired from the from the [payscale.com](https://www.payscale.com). It contains all the schools in United States that are arranged by various measurement of career performance, such as “Early Career Pay” and “Mid Career Pay”.

**Problem encountered** When importing html data from <https://www.payscale.com/college-salary-report/bachelors>, I realized the table only include the data with the top 25 schools in the United States, descending by measurement of career performance. That’s the issue that I am not expecting. Moreover, this is the first page in the web and there are 63 pages in total, which consists all the school data.

**Problem resolved** Instead of importing data 63 times to get the entire dataset, one alternative webpage is found by navigating the [payscale.com](https://www.payscale.com/college-salary-report/best-schools-by-state). The page “Best Schools By State” (<https://www.payscale.com/college-salary-report/best-schools-by-state>) outlays all the best schools ranked by measurement of career performance of all 50 states. Clicking on each state would direct to the schools data within that particular state. In order to import the entire data, I first convert the string format in the list of state.name to match the url (“New York” to “New-York”). Then, a data frame is created. For-Loop is implemented to import all 50 states data to the R environment and to keep loading data into the data frame to form a complete dataset, “`Best_School`”, for data cleansing.

### Data Cleaning for `Best_School` data

First step is to modify the column name “School Name” to “name” and to keep the exact name of school only, in order to match the `tuition_with_diversity` dataset for binding. After that, there are several data cleansing steps that are applied to other columns. Only numeric values are extracted from the column, “Rank”, “Early Career Pay”, “Mid-Career Pay”, “% High Meaning”, “% STEM Degrees”. One lesson learned is that R suggests to use `parse_number()`, instead of `extract_numeric()` for extracting numeric value.

Below is the snippet of Best\_School\_clean data

##	name	Early Career Pay	Mid-Career Pay
## 1	Auburn University	54400	104500
## 2	University of Alabama in Huntsville	57500	103900
## 3	The University of Alabama	52300	97400
## 4	Tuskegee University	54500	93500
## 5	Samford University	48400	90500
## 6	Spring Hill College	46600	89100
## 7	Birmingham Southern College	49100	88300
## 8	University of Alabama at Birmingham	48600	87200
## 9	University of South Alabama	47700	86400
## 10	Alabama A & M University	48700	83500

  

##	% High Meaning	% STEM Degrees
## 1	51	31
## 2	59	45
## 3	50	15
## 4	61	30
## 5	52	3
## 6	53	12
## 7	48	27
## 8	57	17
## 9	56	17
## 10	58	20

Combine Best\_School\_clean data and tuition\_with\_diversity to form the final data

Finally, Best\_School\_clean data, which contains different measurements of career performance, merges with tuition\_with\_diversity data, which contains detailed school information including tuition and race. The column both datasets have in common is “name” and left\_join is performed. Similar to the previous merged dataset, schools with “NA” are removed from the dataset.

Below is the snippet of the Final\_data

There are 622 observations in all 50 states in United States and each college or university is a unique observation. This is the tidy version of the final data and it will be stored as a csv file.

### Attribute Information

Below information is from payscale.com:

“Early Career Pay” is defined as median salary for alumni with 0-5 years experience.

“Mid-Career Pay” is defined as Median salary for alumni with 10+ years experience.

“% High Meaning” is defined as the percentage of alumni who say their work makes the world a better place.

“% STEM Degrees” is defined as the percentage of degrees awarded in science, technology, engineering or a math subjects.

##	name	Early Career Pay	Mid-Career Pay
## 1	Auburn University	54400	104500
## 2	Tuskegee University	54500	93500
## 3	Samford University	48400	90500

## 4	Spring Hill College	46600	89100
## 5	University of Alabama at Birmingham	48600	87200
## 6	University of South Alabama	47700	86400
## 7	Troy University	44500	81500
## 8	Jacksonville State University	43800	80000
## 9	Auburn University at Montgomery	45000	79600
## 10	Huntingdon College	42400	78900
##	% High Meaning % STEM Degrees	state state_code	type degree_length
## 1	51 31 Alabama	AL Public	4 Year
## 2	61 30 Alabama	AL Private	4 Year
## 3	52 3 Alabama	AL Private	4 Year
## 4	53 12 Alabama	AL Private	4 Year
## 5	57 17 Alabama	AL Public	4 Year
## 6	56 17 Alabama	AL Public	4 Year
## 7	60 8 Alabama	AL Public	4 Year
## 8	61 7 Alabama	AL Public	4 Year
## 9	61 12 Alabama	AL Public	4 Year
## 10	69 14 Alabama	AL Private	4 Year
##	in_state_tuition_and_fee out_of_state_tuition_and_fee	ENROLLMENT WOMEN	
## 1	24608 43856	25912 12798	
## 2	31820 31820	3103 1855	
## 3	42200 42200	4933 3082	
## 4	52926 52926	1376 820	
## 5	17110 31030	18698 11288	
## 6	17490 27360	15805 9700	
## 7	20645 31060	19041 11948	
## 8	18525 28245	8659 4978	
## 9	17268 29028	5057 3233	
## 10	37150 37150	1160 572	
##	AMERICAN INDIAN / ALASKA NATIVE ASIAN BLACK HISPANIC		
## 1	183 601 1886 599		
## 2	2 26 2345 32		
## 3	17 80 372 218		
## 4	10 16 210 77		
## 5	46 931 3943 496		
## 6	100 539 3285 402		
## 7	143 140 6840 666		
## 8	61 50 2030 110		
## 9	23 104 1633 36		
## 10	14 9 229 29		
##	NATIVE HAWAIIAN / PACIFIC ISLANDER WHITE TOTAL MINORITY		
## 1	0 20855 3269		
## 2	0 52 2405		
## 3	1 4007 738		
## 4	1 947 359		
## 5	14 11840 5993		
## 6	33 10102 4684		
## 7	19 9265 8294		
## 8	7 5934 2258		
## 9	9 2572 1941		
## 10	2 738 313		

The tidy version of the final data, “Fianl\_data” is saved under the name “Tidy\_Data.xlsx” local location and will be committed from Github desktop to Github.com repository (<https://github.com/Junjie-Dylan-Yang/Data-Wrangling-Project>)

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#### Import fourth data: historical\_tuition

The last data source, historical\_tuition, is from “College tuition, Diversity, and Pay” in [rfordatascience/tidetuesday/2020-03-10](https://rfordatascience.tidetuesday.com/2020-03-10/), which is originally acquired from the National Center for Education Statistics. (<https://nces.ed.gov/fastfacts/display.asp?id=76>)

The fourth data, historical\_tuition, is tidy and contains the information of the trends in the cost of college education. Therefore, “historical\_tuition“ is saved under the name of “Tuition\_trend.xlsx” in the same location of The tidy version of the final data.

Below is the snippet of tuition\_cost data

```
## # A tibble: 10 x 4
##   type          year  tuition_type  tuition_cost
##   <chr>         <chr>   <chr>         <dbl>
## 1 All Institutions 1985-86 All Constant    10893
## 2 All Institutions 1985-86 4 Year Constant  12274
## 3 All Institutions 1985-86 2 Year Constant    7508
## 4 All Institutions 1985-86 All Current    4885
## 5 All Institutions 1985-86 4 Year Current    5504
## 6 All Institutions 1985-86 2 Year Current    3367
## 7 All Institutions 1995-96 All Constant   13822
## 8 All Institutions 1995-96 4 Year Constant   16224
## 9 All Institutions 1995-96 2 Year Constant    7421
## 10 All Institutions 1995-96 All Current    8800
```

### III. Data Analysis by plot and tables

After above series of data wrangling and data cleansing conducted on several data sources, final data in tidy version, “Final\_data” and “historical\_tuition” data are ready to use for data analysis.

#### 1, Tuition Trend