

# Salary\_cleaning

February 21, 2023

```
[211]: # import necessary libraries - (CELL 1)
```

```
import pandas as pd
import matplotlib.pyplot as plt
import numpy as np
from fuzzywuzzy import fuzz
```

```
[212]: # read in the dataset - (CELL 2)
```

```
data = pd.read_csv("Surveys.csv", sep=',').replace('\"', '\', regex=True)
```

```
[213]: # display data to get a grasp on what the whole dataset looks like - (CELL 3)
```

```
display(data.head())
```

	Timestamp	How old are you?	What industry do you work in?	\
0	4/27/2021 11:02:10	25-34	Education (Higher Education)	
1	4/27/2021 11:02:22	25-34	Computing or Tech	
2	4/27/2021 11:02:38	25-34	Accounting, Banking & Finance	
3	4/27/2021 11:02:41	25-34	Nonprofits	
4	4/27/2021 11:02:42	25-34	Accounting, Banking & Finance	

	Job title	\
0	Research and Instruction Librarian	
1	Change & Internal Communications Manager	
2	Marketing Specialist	
3	Program Manager	
4	Accounting Manager	

	If your job title needs additional context, please clarify here:	\
0	NaN	
1	NaN	
2	NaN	
3	NaN	
4	NaN	

	What is your annual salary? (You'll indicate the currency in a later question. ␣ ␣If you are part-time or hourly, please enter an annualized equivalent -- what ␣ ␣you would earn if you worked the job 40 hours a week, 52 weeks a year.) \
0	55,000
1	54,600

2	34,000
3	62,000
4	60,000

How much additional monetary compensation do you get, if any (for example, bonuses or overtime in an average year)? Please only include monetary compensation here, not the value of benefits. \

0	0.0
1	4000.0
2	NaN
3	3000.0
4	7000.0

Please indicate the currency \

0	USD
1	GBP
2	USD
3	USD
4	USD

If "Other," please indicate the currency here: \

0	NaN
1	NaN
2	NaN
3	NaN
4	NaN

If your income needs additional context, please provide it here: \

0	NaN
1	NaN
2	NaN
3	NaN
4	NaN

What country do you work in? \

0	United States
1	United Kingdom
2	US
3	USA
4	US

If you're in the U.S., what state do you work in? What city do you work in? \

0	Massachusetts	Boston
1	NaN	Cambridge
2	Tennessee	Chattanooga
3	Wisconsin	Milwaukee
4	South Carolina	Greenville

```

How many years of professional work experience do you have overall? \
0          5-7 years
1          8 - 10 years
2          2 - 4 years
3          8 - 10 years
4          8 - 10 years

How many years of professional work experience do you have in your field? \
0          5-7 years
1          5-7 years
2          2 - 4 years
3          5-7 years
4          5-7 years

What is your highest level of education completed? What is your gender? \
0          Master's degree          Woman
1          College degree          Non-binary
2          College degree          Woman
3          College degree          Woman
4          College degree          Woman

What is your race? (Choose all that apply.)
0          White
1          White
2          White
3          White
4          White

```

```
[ ]: # (CELL 4)
```

## 1 Data explanation

The dataset consists of survey data from 2021 and 2022 regarding people's salary. It contains 27922 entries and is made out of 18 variables such as salary, job title, industry, etc. The column names are the questions given to the respondent. The survey data can be retrieved from <https://oscarbaruffa.com/messy/>. The survey form can be retrieved from <https://www.askamanager.org/2021/04/how-much-money-do-you-make-4.html>.

The following variables are present in the dataset.

### 1.1 Timestamp

- The datetime at which the respondent submitted their entry.
- This variable was generated by the software handling the survey data
- This variable is never empty.

## **1.2 Age band**

- The age band in which the respondent belongs to.
- The respondent chose from a premade list of answers.
- This variable is mandatory.
- Only 1 answer can be chosen.

## **1.3 Industry**

- The industry in which the respondent works.
- The respondent either chose from a premade list of answers or gave their own answer.
- This variable is not mandatory.
- Only 1 answer could be chosen or given.

## **1.4 Job title**

- The job title of the respondent.
- The respondent gave their own answer.
- This variable is mandatory.
- Only 1 answer can given.

## **1.5 Job context**

- Additional context regarding the respondent's job title.
- The respondent gave their own answer.
- This variable is not mandatory.
- Only 1 answer could be given.

## **1.6 Salary**

- The respondent's annual salary based on 40 hours a week, 52 weeks a year.
- The respondent gave their own answer.
- This variable is mandatory.
- Only 1 answer can be given.

## **1.7 Compensation**

- Additional monetary income if the respondent has any.
- The respondent gave their own answer.
- This variable is not mandatory.
- Only 1 answer could be given.

## **1.8 Currency**

- The currency in which the respondent receives their salary and compensation.
- The respondent chose from a premade list of answers.
- This variable is mandatory ('Other' is an answer in the list).
- Only 1 answer can be chosen.

## **1.9 Other currency**

- The respondent's currency in case it wasn't an option in the premade list.
- The respondent gave their own answer.
- This variable is not mandatory.
- Only 1 answer could be given.

## **1.10 Income context**

- Additional context regarding the salary and compensation of the respondent.
- The respondent gave their own answer.
- This variable is not mandatory.
- Only 1 answer could be given.

## **1.11 Country**

- The country in which the respondent works.
- The respondent gave their own answer.
- This variable is mandatory.
- Only 1 answer can be given.

## **1.12 State**

- The state or states of the respondent in case the respondent works in the USA.
- The respondent chose from a premade list of answers.
- This variable is not mandatory.
- Multiple answers could be chosen.

## **1.13 City**

- The city in which the respondent works.
- The respondent gave their own answer.
- This variable is mandatory.
- Only 1 answer can be given.

### 1.14 Overall professional years of experience band

- The band in which the respondent has overall professional years of experience.
- The respondent chose from a premade list of answers.
- This variable is mandatory.
- Only 1 answer can be chosen.

### 1.15 Field professional years of experience band

- The band in which the respondent professional years of experience in their current field.
- The respondent chose from a premade list of answers.
- This variable is mandatory.
- Only 1 answer can be chosen.

### 1.16 Education

- The respondent's highest level of education.
- The respondent chose from a premade list of answers.
- This variable is not mandatory.
- Only 1 answer could be chosen.

### 1.17 Gender

- The respondent's gender.
- The respondent chose from a premade list of answers ("Other/no answer" is an answer in the list).
- This variable is not mandatory.
- Only 1 answer could be chosen.

### 1.18 Race

- The respondent's race or races.
- The respondent chose from a premade list of answers ("Other/no answer" is an answer in the list).
- This variable is not mandatory.
- Multiple answers could be chosen.

## 2 Data cleansing and preparation

```
[214]: # new column names - (CELL 5)
new_columns = [
    "datetime", "age_band", "industry", "job_title", "job_context", "salary",
    ↪ "compensation", "currency", \
```

```

    "other_currency", "income_context", "country", "state", "city", \
    ↪ "overall_experience_band", "field_experience_band", \
    "education", "gender", "race"
]

# map old column names to new ones
mapping = {}
for i in range(0, len(data.columns)):
    mapping.update({data.columns[i]: new_columns[i]})

# rename columns
data = data.rename(mapping, axis='columns')

# check if columns are renamed
display(data.head())

```

	datetime	age_band	industry	\
0	4/27/2021 11:02:10	25-34	Education (Higher Education)	
1	4/27/2021 11:02:22	25-34	Computing or Tech	
2	4/27/2021 11:02:38	25-34	Accounting, Banking & Finance	
3	4/27/2021 11:02:41	25-34	Nonprofits	
4	4/27/2021 11:02:42	25-34	Accounting, Banking & Finance	

  

	job_title	job_context	salary	compensation	\
0	Research and Instruction Librarian	NaN	55,000	0.0	
1	Change & Internal Communications Manager	NaN	54,600	4000.0	
2	Marketing Specialist	NaN	34,000	NaN	
3	Program Manager	NaN	62,000	3000.0	
4	Accounting Manager	NaN	60,000	7000.0	

  

	currency	other_currency	income_context	country	state	\
0	USD	NaN	NaN	United States	Massachusetts	
1	GBP	NaN	NaN	United Kingdom	NaN	
2	USD	NaN	NaN	US	Tennessee	
3	USD	NaN	NaN	USA	Wisconsin	
4	USD	NaN	NaN	US	South Carolina	

  

	city	overall_experience_band	field_experience_band	education	\
0	Boston	5-7 years	5-7 years	Master's degree	
1	Cambridge	8 - 10 years	5-7 years	College degree	
2	Chattanooga	2 - 4 years	2 - 4 years	College degree	
3	Milwaukee	8 - 10 years	5-7 years	College degree	
4	Greenville	8 - 10 years	5-7 years	College degree	

  

	gender	race
0	Woman	White
1	Non-binary	White

```

2      Woman  White
3      Woman  White
4      Woman  White

```

```

[215]: # check what kind of dtype each column has - (CELL 6)
print(data.info())

```

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 27922 entries, 0 to 27921
Data columns (total 18 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   datetime                             27922 non-null  object
1   age_band                             27922 non-null  object
2   industry                             27850 non-null  object
3   job_title                            27922 non-null  object
4   job_context                           7226 non-null   object
5   salary                               27922 non-null  object
6   compensation                          20677 non-null  float64
7   currency                             27922 non-null  object
8   other_currency                       196 non-null    object
9   income_context                       3033 non-null   object
10  country                              27922 non-null  object
11  state                                22945 non-null  object
12  city                                 27847 non-null  object
13  overall_experience_band               27922 non-null  object
14  field_experience_band                 27922 non-null  object
15  education                            27711 non-null  object
16  gender                               27757 non-null  object
17  race                                 27754 non-null  object
dtypes: float64(1), object(17)
memory usage: 3.8+ MB
None

```

```

[216]: # change datetime format and dtype - (CELL 7)
data['datetime'] = pd.to_datetime(data['datetime'], infer_datetime_format=True)

# check if dtype and values are correct
print(data['datetime'].info())
display(data[['datetime']].head())

```

```

<class 'pandas.core.series.Series'>
RangeIndex: 27922 entries, 0 to 27921
Series name: datetime
Non-Null Count  Dtype
-----
27922 non-null  datetime64[ns]
dtypes: datetime64[ns](1)

```



memory usage: 218.3 KB  
None

```
          datetime
0 2021-04-27 11:02:10
1 2021-04-27 11:02:22
2 2021-04-27 11:02:38
3 2021-04-27 11:02:41
4 2021-04-27 11:02:42
```

```
[217]: # change age band dtype - (CELL 8)
data['age_band'] = data['age_band'].astype('category')

# change industry dtype
data['industry'] = data['industry'].astype('category')

# check if dtype got changed
print(data.info())
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 27922 entries, 0 to 27921
Data columns (total 18 columns):
 #   Column                Non-Null Count  Dtype
---  -
0   datetime              27922 non-null  datetime64[ns]
1   age_band              27922 non-null  category
2   industry              27850 non-null  category
3   job_title             27922 non-null  object
4   job_context           7226 non-null   object
5   salary                27922 non-null  object
6   compensation          20677 non-null  float64
7   currency              27922 non-null  object
8   other_currency        196 non-null    object
9   income_context        3033 non-null   object
10  country               27922 non-null  object
11  state                 22945 non-null  object
12  city                  27847 non-null  object
13  overall_experience_band 27922 non-null  object
14  field_experience_band  27922 non-null  object
15  education              27711 non-null  object
16  gender                 27757 non-null  object
17  race                   27754 non-null  object
dtypes: category(2), datetime64[ns](1), float64(1), object(14)
memory usage: 3.5+ MB
None
```

```
[218]: # check how many NaN industry has - (CELL 9)
print(data['industry'].isnull().sum())
```

```

print(data.shape)

# only take rows with non-NaN
data = data[data['industry'].notna()]

# check if rows are removed
print(data['industry'].isnull().sum())
print(data.shape)

```

```

72
(27922, 18)
0
(27850, 18)

```

```

[219]: # https://towardsdatascience.com/string-matching-with-fuzzywuzzy-e982c61f8a84 -
↳ (CELL 10)
# https://pypi.org/project/fuzzywuzzy/

# replace 'Health care' with 'Health Care'
data['industry'] = data['industry'].replace('Health care', 'Health Care')

# all answers that could be chosen for industry
premade_catgs = [
    'Accounting, Banking & Finance',
    'Agriculture or Forestry',
    'Art & Design',
    'Business or Consulting',
    'Computing or Tech',
    'Education (Primary/Secondary)',
    'Education (Higher Education)',
    'Engineering or Manufacturing',
    'Entertainment',
    'Government and Public Administration',
    'Health Care',
    'Hospitality & Events',
    'Insurance',
    'Law',
    'Law Enforcement & Security',
    'Leisure, Sport & Tourism',
    'Marketing, Advertising & PR',
    'Media & Digital',
    'Nonprofits',
    'Property or Construction',
    'Recruitment or HR',
    'Retail',
    'Sales',
    'Social Work',

```

```

    'Transport or Logistics',
    'Utilities & Telecommunications'
]

# adding this categories manually a good amount of answers involving it
premade_catgs.append('Library or Archiving')

# all answers present in 'industry'
all_catgs = list(data['industry'])

# get all text-input answers
input_catgs = []
for catg in all_catgs:
    if catg not in premade_catgs:
        input_catgs.append(catg)

# map text-input answers to a premade answer
catg_mapping = {}
catg_ratio_threshold = 75
for input_catg in input_catgs:
    matches = []
    for premade_catg in premade_catgs:

        # try to match every text-input answers to a premade answer when a
        ↪ threshold is met
        TSR_score = fuzz.token_set_ratio(input_catg, premade_catg)
        PR_score = fuzz.partial_ratio(input_catg.lower(), premade_catg.lower())
        if TSR_score >= catg_ratio_threshold:
            matches.append(tuple([input_catg, premade_catg, TSR_score]))
        if PR_score >= catg_ratio_threshold:
            matches.append(tuple([input_catg, premade_catg, PR_score]))

    # if a text-input answer has multiple matches, pick the one with the
    ↪ highest score
    if len(matches) > 0:
        best_match = matches[0]
        for match in matches:
            if match[2] > best_match[2]:
                best_match = match
        catg_mapping.update({best_match[0]: best_match[1]})

    # if no match is found, text-input answer is mapped to 'Other'
    else:
        catg_mapping.update({input_catg: "Other"})

```

```

[220]: # replace text-input answers for industry - (CELL 11)
data['industry'] = data['industry'].replace(catg_mapping)

```

```
# check categories present in industry
print(data['industry'].value_counts())
```

Computing or Tech	4671
Education (Higher Education)	2465
Nonprofits	2434
Health Care	1899
Government and Public Administration	1897
Accounting, Banking & Finance	1798
Other	1794
Engineering or Manufacturing	1746
Marketing, Advertising & PR	1123
Law	1097
Business or Consulting	861
Education (Primary/Secondary)	836
Media & Digital	773
Insurance	534
Retail	509
Recruitment or HR	458
Property or Construction	400
Utilities & Telecommunications	374
Art & Design	365
Sales	353
Transport or Logistics	316
Social Work	273
Hospitality & Events	266
Entertainment	253
Agriculture or Forestry	140
Leisure, Sport & Tourism	100
Library or Archiving	68
Law Enforcement & Security	47

Name: industry, dtype: int64

```
[221]: # check how many unique job titles there are before lowercasing - (CELL 12)
print(len(data['job_title'].unique()))

# change job titles to lowercase
data['job_title'] = data['job_title'].str.lower()

# check how many unique job titles there are after lowercasing
print(len(data['job_title'].unique()))
```

14249  
13005

```
[225]: # check how many NaN job context has - (CELL 13)
print(data['job_context'].isnull().sum())

# drop job context, too many NaN's
data = data.drop('job_context', axis='columns')

# check if job context got removed
print(data.shape)
```

```
20636
(27850, 17)
```

```
[226]: # remove american notation - (CELL 14)
data['salary'] = data['salary'].str.replace(',', '')

# change salary dtype
data['salary'] = data['salary'].astype(int)

# check if dtype got changed
print(data.info())
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 27850 entries, 0 to 27921
Data columns (total 17 columns):
#   Column                Non-Null Count  Dtype
---  -
0   datetime              27850 non-null  datetime64[ns]
1   age_band              27850 non-null  category
2   industry              27850 non-null  category
3   job_title             27850 non-null  object
4   salary                27850 non-null  int64
5   compensation          20629 non-null  float64
6   currency              27850 non-null  object
7   other_currency        195 non-null   object
8   income_context        3031 non-null  object
9   country               27850 non-null  object
10  state                 22889 non-null  object
11  city                  27775 non-null  object
12  overall_experience_band 27850 non-null  object
13  field_experience_band  27850 non-null  object
14  education              27642 non-null  object
15  gender                 27686 non-null  object
16  race                   27683 non-null  object
dtypes: category(2), datetime64[ns](1), float64(1), int64(1), object(12)
memory usage: 3.5+ MB
None
```

```
[227]: # check which values are in other currency - (CELL 15)
print(data['other_currency'].value_counts())
```

```
USD
11
INR
10
NOK
10
SGD
9
MYR
8
DKK
8
AUD
7
BRL
6
PLN
5
CZK
4
NZD
4
NTD
2
ILS
2
GBP
2
KRW
2
CNY
2
MXN
2
None
2
ARS
2
Dkk
2
THB
2
IDR
1
```

PHP  
 1  
 RM  
 1  
 Polish Złoty  
 1  
 SAR  
 1  
 Philippine peso (PHP)  
 1  
 Australian Dollars  
 1  
 PhP (Philippine Peso)  
 1  
 Converted mine into USD for your easyness  
 1  
 Israeli Shekels  
 1  
 Many non-salary benefits - travel, free healthcare for self, very low for  
 family, non-taxable housing allowance 1  
 Equity  
 1  
 It's marketed as £22000 but we get paid pro-rats, so no pay for the school  
 holidays. 1  
 additional compensation is for overtime (i am paid hourly) so it varies. i have  
 included an estimate 1  
 Argentinian peso (ARS)  
 1  
 Rs  
 1  
 Argentine Peso  
 1  
 Philippine Pesos  
 1  
 Singapore Dollara  
 1  
 Rupees  
 1  
 5  
 1  
 PLN (Zwoty)  
 1  
 croatian kuna  
 1  
 up to 12% annual bonus  
 1  
 N/a  
 1

Canadian  
 1  
 na  
 1  
 47000  
 1  
 Thai Baht  
 1  
 Option to get 2x or 1.5x if taking on a weekend day in the summer  
 1  
 THAI BAHT  
 1  
 Mexican Pesos  
 1  
 SGD  
 1  
 Euro  
 1  
 dkk  
 1  
 Korean Won  
 1  
 CAD  
 1  
 Danish Kroner  
 1  
 INR (Indian Rupee)  
 1  
 AUD Australian  
 1  
 Ils  
 1  
 LKR  
 1  
 IDR  
 1  
 China RMB  
 1  
 EUR  
 1  
 American Dollars  
 1  
 Additonal = Bonus plus stock  
 1  
 ZAR  
 1  
 RSU / equity  
 1



Bdt

1

Mexican pesos

1

BRL (R\$)

1

Indian rupees

1

TTD

1

COP

1

canadian

1

DKK

1

Base plus Commission

1

SEK

1

BR\$

1

Na

1

KWD

1

CHF

1

0

1

I work for an online state university, managing admissions data. Not direct tech support. 1

My bonus is based on performance up to 10% of salary

1

\$76,302.34

1

Php

1

PLN (Polish zloty)

1

Overtime (about 5 hours a week) and bonus

1

czech crowns

1

6000 in stock grants annually

1

ILS (Shekel)

1

```

Nok
1
Sgd
1
Peso Argentino
1
Czk
1
KRW (Korean Won)
1
Philippine Peso
1
AUD and NZD aren't the same currency, and have absolutely nothing to do with
each other :(
1
Taiwanese dollars
1
RMB (chinese yuan)
1
-
1
NIS (new Israeli shekel)
1
Canadian
1
US Dollar
1
AUD & NZD are not the same currency...
1
55,000
1
ILS/NIS
1
Norwegian kroner (NOK)
1
TRY
1
Stock
1
NGN
1
Name: other_currency, dtype: int64

```

```

[228]: # check how many times 'Other' appears in currency - (CELL 16)
print(data['currency'].value_counts())

```

```

USD      23210
CAD       1660
GBP       1581

```

```

EUR          633
AUD/NZD      498
Other        154
CHF          37
SEK          37
JPY          23
ZAR          13
HKD          4
Name: currency, dtype: int64

```

```

[229]: # drop currency with 'Other' value, too insignificant - (CELL 17)
print(data.shape)
data = data.drop(data[data['currency'] == 'Other'].index)

# drop other currency, too many NaN's
data = data.drop('other_currency', axis='columns')

# check if rows and column got removed
print(data.shape)

```

```

(27850, 17)
(27696, 16)

```

```

[230]: # fill compensation with value '0' if NaN - (CELL 18)
data['compensation'] = data['compensation'].fillna(0)

# change compensation dtype
data['compensation'] = data['compensation'].astype(int)

# check if dtype changed
print(data.info())

```

```

<class 'pandas.core.frame.DataFrame'>
Int64Index: 27696 entries, 0 to 27921
Data columns (total 16 columns):
 #   Column                Non-Null Count  Dtype
---  -
 0   datetime              27696 non-null  datetime64[ns]
 1   age_band              27696 non-null  category
 2   industry              27696 non-null  category
 3   job_title             27696 non-null  object
 4   salary                27696 non-null  int64
 5   compensation          27696 non-null  int64
 6   currency              27696 non-null  object
 7   income_context        3008 non-null   object
 8   country               27696 non-null  object
 9   state                 22876 non-null  object
10  city                  27621 non-null  object

```

```

11 overall_experience_band 27696 non-null object
12 field_experience_band   27696 non-null object
13 education               27491 non-null object
14 gender                  27532 non-null object
15 race                   27532 non-null object
dtypes: category(2), datetime64[ns](1), int64(2), object(11)
memory usage: 3.2+ MB
None

```

```

[231]: # check how many NaN income context has - (CELL 19)
print(data['income_context'].isnull().sum())

# drop income context, too many NaN's
data = data.drop('income_context', axis='columns')

# check if job context got removed
print(data.shape)

```

```

24688
(27696, 15)

```

```

[232]: # CELL 20
display(data.head())

```

```

      datetime age_band      industry \
0 2021-04-27 11:02:10  25-34  Education (Higher Education)
1 2021-04-27 11:02:22  25-34      Computing or Tech
2 2021-04-27 11:02:38  25-34  Accounting, Banking & Finance
3 2021-04-27 11:02:41  25-34      Nonprofits
4 2021-04-27 11:02:42  25-34  Accounting, Banking & Finance

      job_title  salary  compensation  currency \
0  research and instruction librarian  55000      0  USD
1  change & internal communications manager  54600    4000  GBP
2      marketing specialist  34000      0  USD
3      program manager  62000    3000  USD
4      accounting manager  60000    7000  USD

      country      state      city overall_experience_band \
0  United States  Massachusetts  Boston      5-7 years
1  United Kingdom      NaN  Cambridge      8 - 10 years
2      US      Tennessee  Chattanooga      2 - 4 years
3      USA      Wisconsin  Milwaukee      8 - 10 years
4      US  South Carolina  Greenville      8 - 10 years

      field_experience_band      education      gender      race
0      5-7 years  Master's degree      Woman  White
1      5-7 years  College degree  Non-binary  White

```

2	2 - 4 years	College degree	Woman	White
3	5-7 years	College degree	Woman	White
4	5-7 years	College degree	Woman	White

```
[233]: # change country to lowercase - (CELL 21)
data['country'] = data['country'].str.lower()

# remove punctuation
data['country'] = data['country'].str.replace('.', '')

# check which values are present in country
print(data['country'].value_counts())
```

```

united states
9275
usa
8568
us
3333
canada
1588
united states
672
uk
663
united kingdom
563
usa
485
united states of america
440
australia
313
germany
174
england
136
ireland
102
new zealand
99
canada
77
united kingdom
67
france
66
australia
```

66  
united states of america  
47  
netherlands  
46  
spain  
44  
us  
40  
scotland  
39  
uk  
37  
sweden  
34  
belgium  
33  
england  
32  
switzerland  
30  
the netherlands  
29  
japan  
27  
america  
21  
new zealand  
20  
united state  
19  
germany  
19  
ireland  
18  
austria  
17  
finland  
16  
italy  
14  
unites states  
13  
south africa  
13  
netherlands  
11  
denmark

10  
united stated  
10  
switzerland  
7  
israel  
7  
united sates  
6  
sweden  
6  
singapore  
6  
nz  
6  
india  
6  
the united states  
6  
scotland  
6  
england, uk  
5  
greece  
5  
u s  
5  
brazil  
4  
wales  
4  
united state of america  
4  
unitedstates  
4  
romania  
4  
portugal  
4  
latvia  
4  
  
4  
china  
4  
pakistan  
4  
unites states

4  
great britain  
3  
united statea  
3  
norway  
3  
mexico  
3  
isa  
3  
scotland, uk  
3  
south africa  
3  
is  
3  
thailand  
3  
hong kong  
3  
u s  
3  
puerto rico  
3  
unite states  
2  
the netherlands  
2  
remote  
2  
vietnam  
2  
canda  
2  
great britain  
2  
united stares  
2  
estonia  
2  
lithuania  
2  
slovenia  
2  
england, united kingdom  
2  
chile



2  
uk (england)  
2  
northern ireland  
2  
usa tomorrow  
2  
philippines  
2  
poland  
2  
bulgaria  
2  
zimbabwe  
2  
ghana  
2  
kenya  
2  
cyprus  
2  
us  
2  
the us  
2  
colombia  
2  
united sates of america  
2  
ua  
2  
bermuda  
2  
spain  
2  
japan  
2  
luxembourg  
2  
united status  
2  
united kingdom (england)  
2  
belgium  
1  
malaysia  
1  
uganda

1  
england/uk  
1  
san francisco  
1  
united statws  
1  
sri lanka  
1  
ecuador  
1  
malta  
1  
us govt employee overseas, country withheld  
1  
usa-- virgin islands  
1  
contracts  
1  
morocco  
1  
africa  
1  
kuwait  
1  
currently finance  
1  
n/a (remote from wherever i want)  
1  
united stateds  
1  
united sttes  
1  
hungary  
1  
remote (philippines)  
1  
unites kingdom  
1  
global  
1  
nigeria  
1  
panamá  
1  
canada, ottawa, ontario  
1  
austria, but i work remotely for a dutch/british company

1  
i was brought in on this salary to help with the ehr and very quickly was  
promoted to current position but compensation was not altered  
1  
uniter statez  
1  
congo  
1  
uruguay  
1  
britain  
1  
usat  
1  
we don't get raises, we get quarterly bonuses, but they periodically asses  
income in the area you work, so i got a raise because a 3rd party assessment  
showed i was paid too little for the area we were located 1  
luxemburg  
1  
northern ireland  
1  
norway  
1  
jamaica  
1  
usd  
1  
usa, but for foreign gov't  
1  
jordan  
1  
united statss  
1  
i work for a uae-based organization, though i am personally in the us  
1  
united states  
1  
aotearoa new zealand  
1  
na  
1  
policy  
1  
us>  
1  
hong konh  
1  
united states is america

1  
liechtenstein  
1  
company in germany i work from pakistan  
1  
canadá  
1  
united states of american  
1  
australian  
1  
uk, but for globally fully remote company  
1  
california  
1  
ukraine  
1  
unitedf stated  
1  
wales, uk  
1  
united stares  
1  
croatia  
1  
england, united kingdom  
1  
usaa  
1  
south korea  
1  
united states- puerto rico  
1  
europe  
1  
from new zealand but on projects across apac  
1  
y  
1  
united y  
1  
mexico  
1  
wales (uk)  
1  
isle of man  
1  
northern ireland, united kingdom

1  
qatar  
1  
uk, remote  
1  
unitied states  
1  
united states of americas  
1  
united arab emirates  
1  
rwanda  
1  
uk (northern ireland)  
1  
uk for us company  
1  
us of a  
1  
hong kong  
1  
canad  
1  
uniyes states  
1  
eritrea  
1  
uniyed states  
1  
cambodia  
1  
i am located in canada but i work for a company in the us  
1  
can  
1  
cayman islands  
1  
bangladesh  
1  
united statees  
1  
csnada  
1  
japan, us gov position  
1  
hartford  
1  
new zealand aotearoa

1  
serbia  
1  
russia  
1  
uxz  
1  
united kindom  
1  
puerto rico  
1  
canada and usa  
1  
catalonia  
1  
\$2,17584/year is deducted for benefits  
1  
france  
1  
italy (south)  
1  
jersey, channel islands  
1  
virginia  
1  
afghanistan  
1  
uss  
1  
uniteed states  
1  
united stattes  
1  
for the united states government, but posted overseas  
1  
usab  
1  
worldwide (based in us but short term trips aroundn the world)  
1  
englang  
1  
united statew  
1  
uae  
1  
canadw  
1  
bonus based on meeting yearly goals set w/ my supervisor

1  
international  
1  
the bahamas  
1  
i earn commission on sales if i meet quota, i'm guaranteed another 16k min last  
year i earned an additional 27k it's not uncommon for people in my space to earn  
100k+ after commission 1  
united statesp  
1  
costa rica  
1  
united states  
1  
united statues  
1  
argentina  
1  
untied states  
1  
uniited states  
1  
united states of american  
1  
sierra leone  
1  
portugal  
1  
slovakia  
1  
nederland  
1  
united kingdomk  
1  
unted states  
1  
new zealand  
1  
cuba  
1  
united states (i work from home and my clients are all over the us/canada/pr  
1  
australi  
1  
cote d'ivoire  
1  
from romania, but for an us based company  
1

```

somalia
1
wales (united kingdom)
1
england, gb
1
danmark
1
uk (northern england)
1
malaysia
1
nl
1
bosnia and herzegovina
1
Name: country, dtype: int64

```

/tmp/ipykernel\_41508/1744000476.py:5: FutureWarning: The default value of regex will change from True to False in a future version. In addition, single character regular expressions will *not* be treated as literal strings when regex=True.

```
data['country'] = data['country'].str.replace('.', '')
```

[234]: *# list with correct values - (CELL 22)*

```

correct_countries = [
    'united states of america',
    'united states',
    'usa',
    'united kingdom',
    'great britain'
    'uk',
    'england',
    'canada',
    'germany',
    'france',
    'spain',
    'scotland',
    'netherlands',
    'australia',
    'austria',
    'new zealand',
    'argentina',
    'italy',
    'finland',
    'wales',
    'ireland',
    'belgium',

```



```

    'switzerland',
    'japan',
    'south africa',
    'denmark'
]

input_countries = list(data['country'])

# map incorrect country to correct country
country_mapping = {}
country_ratio_threshold = 75
for input_country in input_countries:
    matches = []
    for correct_country in correct_countries:

        # try to match every text-input answers to a correct country variation
        ↪when a threshold is met
        TSR_score = fuzz.token_set_ratio(input_country, correct_country)
        if TSR_score >= country_ratio_threshold:
            matches.append(tuple([input_country, correct_country, TSR_score]))

        # if a text-input answer has multiple matches, pick the one with the
        ↪highest score
        if len(matches) > 0:
            best_match = matches[0]
            for match in matches:
                if match[2] > best_match[2]:
                    best_match = match
            country_mapping.update({best_match[0]: best_match[1]})

        # if no match is found, text-input answer is mapped to 'unknown'
        else:
            country_mapping.update({input_catg: "unknown"})

```

```

[235]: # replace text-input answers for country - (CELL 23)
data['country'] = data['country'].replace(country_mapping)

# remove trailing whitespaces
data['country'] = data['country'].str.strip()

# check categories present in country
print(data['country'].value_counts())

```

```

usa
12439
united states of america
10464

```

canada  
1673  
uk  
700  
united kingdom  
640  
australia  
381  
germany  
194  
england  
179  
ireland  
124  
new zealand  
123  
united states  
93  
netherlands  
89  
france  
67  
scotland  
48  
spain  
46  
sweden  
40  
switzerland  
37  
belgium  
34  
japan  
30  
austria  
18  
south africa  
17  
finland  
16  
italy  
15  
denmark  
11  
u s  
8  
israel  
7

singapore  
6  
wales  
6  
india  
6  
nz  
6  
greece  
5  
great britainuk  
5  
portugal  
5  
puerto rico  
4  
brazil  
4  
mexico  
4  
romania  
4  
norway  
4  
latvia  
4  
unitedstates  
4  
pakistan  
4  
china  
4  
  
4  
hong kong  
4  
isa  
3  
thailand  
3  
is  
3  
slovenia  
2  
vietnam  
2  
remote  
2

cyprus  
 2  
 malaysia  
 2  
 chile  
 2  
 philippines  
 2  
 poland  
 2  
 bulgaria  
 2  
 estonia  
 2  
 zimbabwe  
 2  
 ghana  
 2  
 kenya  
 2  
 the us  
 2  
 lithuania  
 2  
 bermuda  
 2  
 colombia  
 2  
 luxembourg  
 2  
 uk for us company  
 1  
 global  
 1  
 contracts  
 1  
 san francisco  
 1  
 we don't get raises, we get quarterly bonuses, but they periodically asses  
 income in the area you work, so i got a raise because a 3rd party assessment  
 showed i was paid too little for the area we were located 1  
 britain  
 1  
 ecuador  
 1  
 morocco  
 1  
 malta

1  
worldwide (based in us but short term trips around the world)  
1  
croatia  
1  
uganda  
1  
us govt employee overseas, country withheld  
1  
n/a (remote from wherever i want)  
1  
cayman islands  
1  
uruguay  
1  
luxemburg  
1  
south korea  
1  
hartford  
1  
ukraine  
1  
liechtenstein  
1  
hong kong  
1  
policy  
1  
na  
1  
i work for a uae-based organization, though i am personally in the us  
1  
jordan  
1  
kuwait  
1  
usd  
1  
jamaica  
1  
uk, but for globally fully remote company  
1  
california  
1  
europe  
1  
isle of man

1  
sri lanka  
1  
y  
1  
congo  
1  
i was brought in on this salary to help with the ehr and very quickly was  
promoted to current position but compensation was not altered  
1  
can  
1  
catalonia  
1  
serbia  
1  
russia  
1  
somalia  
1  
from romania, but for an us based company  
1  
cote d'ivoire  
1  
uxz  
1  
cuba  
1  
eritrea  
1  
panamá  
1  
\$2,17584/year is deducted for benefits  
1  
jersey, channel islands  
1  
bangladesh  
1  
virginia  
1  
afghanistan  
1  
uss  
1  
cambodia  
1  
nl  
1

```

slovakia
1
currently finance
1
sierra leone
1
nigeria
1
remote (philippines)
1
hungary
1
qatar
1
uk, remote
1
us of a
1
argentina
1
costa rica
1
i earn commission on sales if i meet quota, i'm guaranteed another 16k min last
year i earned an additional 27k it's not uncommon for people in my space to earn
100k+ after commission 1
rwanda
1
the bahamas
1
international
1
bonus based on meeting yearly goals set w/ my supervisor
1
united arab emirates
1
uae
1
bosnia and herzegovina
1
Name: country, dtype: int64

```

```

[236]: # improve certain mappings - (CELL 24)
improved_country_mapping = {
    'usa': 'united states of america',
    'united states': 'united states of america',
    'uk': 'united kingdom',
    'england': 'united kingdom',

```

```

    'great britain': 'united kingdom',
    'scotland': 'united kingdom',
    'wales': 'united kingdom'
}

# refine mappings for country
data['country'] = data['country'].replace(improved_country_mapping)

# check categories present in country
print(data['country'].value_counts())

```

```

united states of america
22996
canada
1673
united kingdom
1573
australia
381
germany
194
ireland
124
new zealand
123
netherlands
89
france
67
spain
46
sweden
40
switzerland
37
belgium
34
japan
30
austria
18
south africa
17
finland
16
italy
15

```



denmark  
11  
u s  
8  
israel  
7  
nz  
6  
singapore  
6  
india  
6  
greece  
5  
portugal  
5  
great britainuk  
5  
pakistan  
4  
  
4  
unitedstates  
4  
china  
4  
latvia  
4  
romania  
4  
brazil  
4  
puerto rico  
4  
norway  
4  
mexico  
4  
hong kong  
4  
is  
3  
isa  
3  
thailand  
3  
philippines  
2

zimbabwe  
2  
ghana  
2  
lithuania  
2  
estonia  
2  
bulgaria  
2  
remote  
2  
vietnam  
2  
malaysia  
2  
the us  
2  
poland  
2  
slovenia  
2  
chile  
2  
luxembourg  
2  
bermuda  
2  
kenya  
2  
cyprus  
2  
colombia  
2  
ecuador  
1  
i was brought in on this salary to help with the ehr and very quickly was  
promoted to current position but compensation was not altered  
1  
panamá  
1  
morocco  
1  
uruguay  
1  
congo  
1  
uganda

1  
malta  
1  
n/a (remote from wherever i want)  
1  
us govt employee overseas, country withheld  
1  
can  
1  
luxemburg  
1  
san francisco  
1  
jamaica  
1  
liechtenstein  
1  
hong konh  
1  
policy  
1  
na  
1  
i work for a uae-based organization, though i am personally in the us  
1  
jordan  
1  
usd  
1  
uk, but for globally fully remote company  
1  
remote (philippines)  
1  
california  
1  
europe  
1  
isle of man  
1  
y  
1  
ukraine  
1  
south korea  
1  
croatia  
1  
nigeria

1  
costa rica  
1  
hungary  
1  
jersey, channel islands  
1  
cambodia  
1  
global  
1  
worldwide (based in us but short term trips around the world)  
1  
uk for us company  
1  
hartford  
1  
uss  
1  
afghanistan  
1  
virginia  
1  
\$2,17584/year is deducted for benefits  
1  
bangladesh  
1  
catalonia  
1  
uxz  
1  
russia  
1  
serbia  
1  
currently finance  
1  
united arab emirates  
1  
rwanda  
1  
us of a  
1  
britain  
1  
we don't get raises, we get quarterly bonuses, but they periodically assess  
income in the area you work, so i got a raise because a 3rd party assessment  
showed i was paid too little for the area we were located 1

```

qatar
1
sierra leone
1
uk, remote
1
argentina
1
cayman islands
1
i earn commission on sales if i meet quota, i'm guaranteed another 16k min last
year i earned an additional 27k it's not uncommon for people in my space to earn
100k+ after commission
1
the bahamas
1
international
1
bonus based on meeting yearly goals set w/ my supervisor
1
uae
1
kuwait
1
eritrea
1
slovakia
1
nl
1
sri lanka
1
somalia
1
from romania, but for an us based company
1
cote d'ivoire
1
cuba
1
contracts
1
bosnia and herzegovina
1
Name: country, dtype: int64

```

```
[237]: # create df of correct countries - (CELL 25)
```

```
correct_countries_df = pd.DataFrame(correct_countries).rename({0: 'country'},
    ↪axis='columns')
display(correct_countries_df.head())
```

```

          country
0  united states of america
1          united states
2              usa
3    united kingdom
4    great britainuk
```

```
[238]: # create filter to remove remaining countries - (CELL 26)
print(data.shape)
is_correct_country = data['country'].isin(correct_countries_df['country'])
data = data.drop(data[~is_correct_country].index)

# check if rows got dropped
print(data.shape)
```

```
(27696, 15)
```

```
(27450, 15)
```

```
[239]: # change country dtype - (CELL 27)
data['country'] = data['country'].astype('category')

# check if dtype got changed
print(data.info())
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 27450 entries, 0 to 27921
Data columns (total 15 columns):
#   Column                Non-Null Count  Dtype
---  -
0   datetime              27450 non-null  datetime64[ns]
1   age_band              27450 non-null  category
2   industry              27450 non-null  category
3   job_title             27450 non-null  object
4   salary                27450 non-null  int64
5   compensation          27450 non-null  int64
6   currency              27450 non-null  object
7   country               27450 non-null  category
8   state                 22831 non-null  object
9   city                  27376 non-null  object
10  overall_experience_band 27450 non-null  object
11  field_experience_band   27450 non-null  object
12  education              27248 non-null  object
13  gender                 27289 non-null  object
14  race                   27289 non-null  object
```

```
dtypes: category(3), datetime64[ns](1), int64(2), object(9)
memory usage: 2.8+ MB
None
```

```
[240]: # create filter to drop 'united states of america' with no state - (CELL 28)
print(data.shape)
usa_no_state = (data['country'] == 'united states of america') & (data['state'].
↳ isna())

# drop rows
data = data.drop(data[usa_no_state].index)

# check if rows got dropped
print(data.shape)
```

```
(27450, 15)
```

```
(27280, 15)
```

```
[241]: # fill state NaN with value 'Not American' - (CELL 29)
data['state'] = data['state'].fillna('Not American')

# check if NaN still exists
print(data.info())
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 27280 entries, 0 to 27921
Data columns (total 15 columns):
 #   Column                                Non-Null Count  Dtype
---  -
 0   datetime                             27280 non-null  datetime64[ns]
 1   age_band                             27280 non-null  category
 2   industry                             27280 non-null  category
 3   job_title                            27280 non-null  object
 4   salary                               27280 non-null  int64
 5   compensation                         27280 non-null  int64
 6   currency                             27280 non-null  object
 7   country                              27280 non-null  category
 8   state                                27280 non-null  object
 9   city                                 27210 non-null  object
10  overall_experience_band               27280 non-null  object
11  field_experience_band                 27280 non-null  object
12  education                            27090 non-null  object
13  gender                               27127 non-null  object
14  race                                 27132 non-null  object
dtypes: category(3), datetime64[ns](1), int64(2), object(9)
memory usage: 2.8+ MB
None
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 27280 entries, 0 to 27921
Data columns (total 15 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   datetime                             27280 non-null  datetime64[ns]
1   age_band                             27280 non-null  category
2   industry                             27280 non-null  category
3   job_title                            27280 non-null  object
4   salary                               27280 non-null  int64
5   compensation                         27280 non-null  int64
6   currency                             27280 non-null  object
7   country                              27280 non-null  category
8   state                                27280 non-null  category
9   city                                 27210 non-null  object
10  overall_experience_band               27280 non-null  object
11  field_experience_band                 27280 non-null  object
12  education                            27090 non-null  object
13  gender                               27127 non-null  object
14  race                                 27132 non-null  object
dtypes: category(4), datetime64[ns](1), int64(2), object(8)
memory usage: 2.6+ MB
None
```

(27280, 15)  
(27280, 14)

48



```
data['field_experience_band'] = data['field_experience_band'].str.replace(' -',  
    ↪', '-')
```

```
# check if typos got fixed  
print(data['overall_experience_band'].value_counts(), '\n')  
print(data['overall_experience_band'].value_counts())
```

```
11-20 years      9380  
8-10 years       5264  
5-7 years        4739  
21-30 years      3547  
2-4 years        2892  
31-40 years       846  
1 year or less   493  
41 years or more  119  
Name: overall_experience_band, dtype: int64
```

```
11-20 years      9380  
8-10 years       5264  
5-7 years        4739  
21-30 years      3547  
2-4 years        2892  
31-40 years       846  
1 year or less   493  
41 years or more  119  
Name: overall_experience_band, dtype: int64
```

```
[245]: # change overall experience band and field experience band to category - (CELL  
    ↪33)  
data['overall_experience_band'] = data['overall_experience_band'].  
    ↪astype('category')  
data['field_experience_band'] = data['field_experience_band'].astype('category')  
  
# check if dtypes got changed  
print(data.info())
```

```
<class 'pandas.core.frame.DataFrame'>  
Int64Index: 27280 entries, 0 to 27921  
Data columns (total 14 columns):  
#   Column              Non-Null Count  Dtype  
---  ---  
0   datetime            27280 non-null  datetime64[ns]  
1   age_band            27280 non-null  category  
2   industry            27280 non-null  category  
3   job_title           27280 non-null  object  
4   salary              27280 non-null  int64  
5   compensation        27280 non-null  int64
```

```

6   currency                27280 non-null object
7   country                 27280 non-null category
8   state                   27280 non-null category
9   overall_experience_band 27280 non-null category
10  field_experience_band   27280 non-null category
11  education               27090 non-null object
12  gender                  27127 non-null object
13  race                    27132 non-null object
dtypes: category(6), datetime64[ns](1), int64(2), object(5)
memory usage: 2.1+ MB
None

```

```

[249]: # check values for education - (CELL 34)
print(data['education'].value_counts())

```

```

College degree                13178
Master's degree               8630
Some college                  1995
PhD                           1383
Professional degree (MD, JD, etc.) 1295
High School                   609
Name: education, dtype: int64

```

```

[248]: # check how many NaN's education has - (CELL 35)
print(data.shape)
print(data['education'].isnull().sum())

# drop rows with NaN's (no viable way to fill)
data = data[data['education'].notna()]

# check if rows got dropped
print(data.shape)

```

```

(27280, 14)
190
(27090, 14)

```

```

[250]: # change education dtype - (CELL 36)
data['education'] = data['education'].astype('category')

# check if dtype has changed
print(data.info())

```

```

<class 'pandas.core.frame.DataFrame'>
Int64Index: 27090 entries, 0 to 27921
Data columns (total 14 columns):
#   Column                Non-Null Count  Dtype
---
```

```

0    datetime                27090 non-null datetime64[ns]
1    age_band                27090 non-null category
2    industry                27090 non-null category
3    job_title               27090 non-null object
4    salary                  27090 non-null int64
5    compensation            27090 non-null int64
6    currency                27090 non-null object
7    country                 27090 non-null category
8    state                   27090 non-null category
9    overall_experience_band  27090 non-null category
10   field_experience_band   27090 non-null category
11   education               27090 non-null category
12   gender                  26966 non-null object
13   race                    26974 non-null object
dtypes: category(7), datetime64[ns](1), int64(2), object(4)
memory usage: 1.9+ MB
None

```

```

[252]: # check values for gender - (CELL 37)
print(data['gender'].value_counts())

```

```

Woman                20732
Man                  5225
Non-binary           734
Other or prefer not to answer 274
Prefer not to answer 1
Name: gender, dtype: int64

```

```

[254]: # remap some gender values - (CELL 38)
gender_mapping = {
    'Other or prefer not to answer': 'No answer',
    'Prefer not to answer': 'No answer'
}

# replace values
data['gender'] = data['gender'].replace(gender_mapping)

# check if values got replaced
print(data['gender'].value_counts())

```

```

Woman                20732
Man                  5225
Non-binary           734
No answer            275
Name: gender, dtype: int64

```

```
[257]: # check for NaN's in gender - (CELL 39)
print(data['gender'].isnull().sum())

# fill NaN's with 'No answer'
data['gender'] = data['gender'].fillna('No answer')

# change gender type
data['gender'] = data['gender'].astype('category')

# check to see if NaN's are gone and dtype is changed
print(data.info())
```

```
0
<class 'pandas.core.frame.DataFrame'>
Int64Index: 27090 entries, 0 to 27921
Data columns (total 14 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   datetime                             27090 non-null  datetime64[ns]
1   age_band                             27090 non-null  category
2   industry                             27090 non-null  category
3   job_title                            27090 non-null  object
4   salary                               27090 non-null  int64
5   compensation                         27090 non-null  int64
6   currency                             27090 non-null  object
7   country                             27090 non-null  category
8   state                                27090 non-null  category
9   overall_experience_band              27090 non-null  category
10  field_experience_band                27090 non-null  category
11  education                            27090 non-null  category
12  gender                               27090 non-null  category
13  race                                 26974 non-null  object
dtypes: category(8), datetime64[ns](1), int64(2), object(3)
memory usage: 1.7+ MB
None
```

```
[261]: # drop race (prefered to not use this data for predictive modeling) - (CELL 40)
data = data.drop('race', axis='columns')

# check if column got dropped
print(data.info())
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 27090 entries, 0 to 27921
Data columns (total 13 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   datetime                             27090 non-null  datetime64[ns]
```

```
1  age_band                27090 non-null  category
2  industry                27090 non-null  category
3  job_title               27090 non-null  object
4  salary                  27090 non-null  int64
5  compensation            27090 non-null  int64
6  currency                27090 non-null  object
7  country                 27090 non-null  category
8  state                   27090 non-null  category
9  overall_experience_band  27090 non-null  category
10 field_experience_band    27090 non-null  category
11 education                27090 non-null  category
12 gender                   27090 non-null  category
dtypes: category(8), datetime64[ns](1), int64(2), object(2)
memory usage: 1.5+ MB
None
```

```
[266]: # save cleaned data - (CELL 41)
data.to_csv('Surveys_cleaned.csv', index=False)
```