analysis

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Description: The following python codes are written for an application to University of Chicago M.S. Applied Data Science program.
1. I shall write a function to randomly select 20% from `bank-additional-full.csv` to form `.csv`. 2. I want to I want to perform some data transformation and standarization. - Add a new column for age range - remove hidden characters from `job` column - Create dict in `Education` - Treat `unknown` value as None type for columns `Default`, `Housing`, `Loan`. - Transform columns to binary `Default`, `Housing`, `Loan`, `Y` - Calculate basic stats: - overall martial status, success rate, job categories, education level, house - Age distribution (visualization) colored by marital status - Job and housing

Moro, S., P. Rita, and P. Cortez. 2014. Bank Marketing. UCI Machine Learning Repository. https://doi.org/10.24432/C5K306.

[1]:
```python
import csv, random
import pandas as pd
import numpy as np

class Utility:
    def __init__(self):
        pass

    def random_select(self, source: str, target: str, size: float) -> str:
        with open(source, 'r') as file:
            reader = csv.reader(file, delimiter= ';')
            header = next(reader)
            data = list(reader)
            size = round(size * len(data))
            sample_rows = random.sample(data, min(size, len(data)))

        with open(target, 'w', newline= '') as outfile:
            writer = csv.writer(outfile, delimiter= ',')
            writer.writerow(header)
            writer.writerows(sample_rows)

        return target
```

```
    def formatEducation(self, value):
        pass

    def transformBinary(self, df, columns: list):
        mapping = {
            'yes': True,
            'no': False,
            'unknown': np.nan
        }
        for column in columns:
            df[column] = df[column].replace(mapping)

        return df

    def AgeMasking(self, age: int):
        age_gorup = np.nan
        if age < 18:
            age_gorup = '< 18'
        elif 18 <= age < 37:
            age_gorup = '18-36'
        elif 37 <= age < 55:
            age_gorup = '37-54'
        elif 54 <= age < 73:
            age_gorup = '54-72'
        elif 73 <= age < 91:
            age_gorup = '73-90'
        elif age >= 90:
            age_gorup = '90+'
        return age_gorup
```

[2]:
```
source='bank-additional-full.csv'
target='bank-sample.csv'
size = 0.1
with open(source, 'r') as file:
    reader = csv.reader(file, delimiter= ';')
    header = next(reader)
    data = list(reader)
    size = round(size * len(data))
    sample_rows = random.sample(data, min(size, len(data)))

with open(target, 'w', newline= '') as outfile:
    writer = csv.writer(outfile, delimiter= ',')
    writer.writerow(header)
    writer.writerows(sample_rows)
```

[3]:
```
import pandas as pd
```

```
util = Utility()
sample = util.random_select(source='bank-additional-full.csv',␣
 ↪target='bank-sample.csv', size=0.15)
data = pd.read_csv(sample)
data.head()
```

[3]:
```
   age            job   marital            education default housing loan  \
0   38     technician   married  professional.course      no     yes   no
1   36        admin.  divorced   university.degree      no     yes   no
2   37       services  divorced            basic.9y      no      no   no
3   44  self-employed   married            basic.9y      no     yes   no
4   45    blue-collar  divorced            basic.9y      no      no   no

    contact month day_of_week  …  campaign  pdays  previous     poutcome  \
0  cellular   mar         fri  …         5    999         0  nonexistent
1  cellular   aug         mon  …         2    999         2      failure
2  cellular   jul         wed  …         1    999         0  nonexistent
3  cellular   nov         thu  …         4    999         1      failure
4  cellular   apr         fri  …         3    999         0  nonexistent

  emp.var.rate  cons.price.idx  cons.conf.idx  euribor3m  nr.employed    y
0         -1.8          92.843          -50.0      1.726       5099.1   no
1         -2.9          92.201          -31.4      0.884       5076.2  yes
2          1.4          93.918          -42.7      4.957       5228.1   no
3         -0.1          93.200          -42.0      4.076       5195.8   no
4         -1.8          93.075          -47.1      1.405       5099.1   no

[5 rows x 21 columns]
```

[4]:
```
data['job'].apply(lambda val: val.rstrip(".,!?"))
```

[4]:
```
0          technician
1               admin
2            services
3       self-employed
4         blue-collar
             …
6173            admin
6174      blue-collar
6175       technician
6176            admin
6177      blue-collar
Name: job, Length: 6178, dtype: object
```

[5]:
```
data["education"].apply(lambda val: val.replace('.', '-'))
```

[5]:
```
0       professional-course
1         university-degree
2                   basic-9y
3                   basic-9y
4                   basic-9y
               …
6173            high-school
6174               basic-4y
6175    professional-course
6176            high-school
6177               basic-9y
Name: education, Length: 6178, dtype: object
```

[7]:
```
new_data = util.transformBinary(df = data, columns = ['default',␣
 ↪'housing','loan','y'])
new_data
```

[7]:

	age	job	marital	education	default	housing	loan	contact	month	day_of_week	…	campaign	pdays	previous	poutcome	emp.var.rate	cons.price.idx	cons.conf.idx	euribor3m	\
0	38	technician	married	professional.course	False	True	False	cellular	mar	fri	…	5	999	0	nonexistent	-1.8	92.843	-50.0	1.726	
1	36	admin.	divorced	university.degree	False	True	False	cellular	aug	mon	…	2	999	2	failure	-2.9	92.201	-31.4	0.884	
2	37	services	divorced	basic.9y	False	False	False	cellular	jul	wed	…	1	999	0	nonexistent	1.4	93.918	-42.7	4.957	
3	44	self-employed	married	basic.9y	False	True	False	cellular	nov	thu	…	4	999	1						
4	45	blue-collar	divorced	basic.9y	False	False	False	cellular	apr	fri	…	3	999	0						
…	…	…	…	…	…	…	…	…	…	…	…	…	…	…						
6173	58	admin.	divorced	high.school	False	False	False	telephone	oct	thu	…	1	999	0						
6174	34	blue-collar	married	basic.4y	False	True	False	telephone	may	tue	…	2	999	0						
6175	34	technician	married	professional.course	False	NaN	NaN	cellular	may	thu	…	3	999	0						
6176	37	admin.	married	high.school	False	False	True	cellular	nov	mon	…	3	999	0						
6177	32	blue-collar	married	basic.9y	False	True	False	telephone	may	fri	…	1	999	0						

```
3         failure         -0.1          93.200          -42.0      4.076
4     nonexistent         -1.8          93.075          -47.1      1.405
…             …            …              …              …          …
6173  nonexistent         -0.1          93.798          -40.4      4.794
6174  nonexistent          1.1          93.994          -36.4      4.857
6175  nonexistent         -1.8          92.893          -46.2      1.327
6176  nonexistent         -3.4          92.649          -30.1      0.722
6177  nonexistent          1.1          93.994          -36.4      4.855

      nr.employed      y
0          5099.1  False
1          5076.2   True
2          5228.1  False
3          5195.8  False
4          5099.1  False
…             …      …
6173       5195.8  False
6174       5191.0  False
6175       5099.1  False
6176       5017.5  False
6177       5191.0  False

[6178 rows x 21 columns]
```

[8]:
```
new_data['age_group']=new_data['age'].apply(lambda val: util.AgeMasking(val))
```

[9]:
```
new_data['age'].describe()
```

[9]:
```
count    6178.000000
mean       39.825834
std        10.461371
min        18.000000
25%        32.000000
50%        38.000000
75%        47.000000
max        98.000000
Name: age, dtype: float64
```

Calculate stats

[11]:
```
new_data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 6178 entries, 0 to 6177
Data columns (total 22 columns):
 #   Column          Non-Null Count  Dtype
---  ------          --------------  -----
 0   age             6178 non-null   int64
```

```
 1   job             6178 non-null   object
 2   marital         6178 non-null   object
 3   education       6178 non-null   object
 4   default         4890 non-null   object
 5   housing         6026 non-null   object
 6   loan            6026 non-null   object
 7   contact         6178 non-null   object
 8   month           6178 non-null   object
 9   day_of_week     6178 non-null   object
 10  duration        6178 non-null   int64
 11  campaign        6178 non-null   int64
 12  pdays           6178 non-null   int64
 13  previous        6178 non-null   int64
 14  poutcome        6178 non-null   object
 15  emp.var.rate    6178 non-null   float64
 16  cons.price.idx  6178 non-null   float64
 17  cons.conf.idx   6178 non-null   float64
 18  euribor3m       6178 non-null   float64
 19  nr.employed     6178 non-null   float64
 20  y               6178 non-null   bool
 21  age_group       6178 non-null   object
dtypes: bool(1), float64(5), int64(5), object(11)
memory usage: 1019.7+ KB
```

[12]:
```
new_data.pivot_table(
    values='duration',
    index=['education', 'housing'],
    columns=['marital'],
    aggfunc='mean',
    fill_value="Null"
    )
```

[12]:

marital		divorced	married	single	unknown
education	housing				
basic.4y	False	173.827586	248.829787	287.0	Null
	True	317.864865	276.590717	180.0	Null
basic.6y	False	427.285714	260.942029	349.166667	Null
	True	214.777778	264.066667	323.318182	Null
basic.9y	False	264.333333	252.69863	249.063158	Null
	True	241.38	253.84984	260.612613	139.0
high.school	False	203.255556	283.79697	257.248908	95.0
	True	251.25	245.051345	264.927419	Null
illiterate	False	146.0	Null	Null	Null
	True	Null	51.0	259.0	Null
professional.course	False	217.057143	240.297674	298.695238	Null
	True	316.189655	259.964567	186.789855	Null
university.degree	False	249.340909	242.041284	243.841085	152.5

```
                    True     304.139535  225.542533       247.6    251.5
unknown             False    143.222222  227.236842    293.9375    170.0
                    True        202.75   289.457831  331.181818     49.0
```

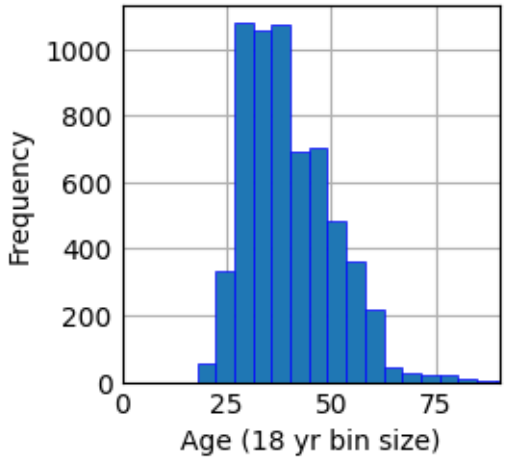
[72]:
```
new_data['age_group']=new_data['age'].apply(lambda val: util.AgeMasking(val))
```

[25]:
```
import matplotlib.pyplot as plt

fig, ax = plt.subplots()

ax.hist(new_data['age'], bins = 18, linewidth = 0.5, edgecolor = 'blue')
ax.set(xlim=(0, 90), xlabel= "Age (18 yr bin size)", ylabel= "Frequency")

plt.show()
```



[ ]:
```
plt.style.use('_mpl-gallery')
bar_fig, bar_ax = plt.subplot()
bar_ax.bar(new_data['job'], new_data, edgecolor = 'white', linewidth = 0.5)
plt.show()
```

```
---------------------------------------------------------------------------
TypeError                                 Traceback (most recent call last)
Cell In[29], line 2
      1 plt.style.use('_mpl-gallery')
----> 2 bar_fig, bar_ax = plt.subplot()
      3 bar_ax.bar(new_data['job'].to_list(), new_data, edgecolor = 'white',␣
 ↪linewidth = 0.5)
      4 plt.show()
```

```
TypeError: cannot unpack non-iterable Axes object
```