

## ✓ Age Vs Salary Classification either above 50k or less 50k through logistic regression classification

Reference of data set: <https://www.kaggle.com/wenrui/Adult-Income-Dataset>



```
1 import pandas as pd
2 import matplotlib.pyplot as plt
3 import numpy as np
4 %matplotlib inline
```

```
1 df = pd.read_csv(r'https://github.com/kaopanboonyuen/Python-Data-Science/raw/master/Dataset/a
2 df.head()
```

	age	workclass	fnlwgt	education	educational-num	marital-status	occupation	relat
0	25	Private	226802	11th	7	Never-married	Machine-op-inspct	
1	38	Private	89814	HS-grad	9	Married-civ-spouse	Farming-fishing	
				Assoc-		Married-	Protective-	

```
1 df.columns
```

```
Index(['age', 'workclass', 'fnlwgt', 'education', 'educational-num',
       'marital-status', 'occupation', 'relationship', 'race', 'gender',
       'capital-gain', 'capital-loss', 'hours-per-week', 'native-country',
       'income'],
      dtype='object')
```

```
1 df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 48842 entries, 0 to 48841
Data columns (total 15 columns):
#   Column                Non-Null Count  Dtype
---  -
0   age                   48842 non-null  int64
1   workclass             48842 non-null  object
2   fnlwgt                48842 non-null  int64
3   education             48842 non-null  object
4   educational-num       48842 non-null  int64
5   marital-status       48842 non-null  object
6   occupation            48842 non-null  object
7   relationship          48842 non-null  object
8   race                  48842 non-null  object
9   gender                48842 non-null  object
10  capital-gain          48842 non-null  int64
11  capital-loss          48842 non-null  int64
12  hours-per-week        48842 non-null  int64
13  native-country        48842 non-null  object
14  income                48842 non-null  object
dtypes: int64(6), object(9)
memory usage: 5.6+ MB
```

## ✓ Analysing data

```
1 df['income'].value_counts()
```

```
<=50K    37155
>50K     11687
Name: income, dtype: int64
```

```
1 df['age'].value_counts()
```

```
36    1348
35    1337
33    1335
23    1329
31    1325
...
88      6
85      5
87      3
89      2
86      1
Name: age, Length: 74, dtype: int64
```

```
1 df['workclass'].value_counts()
```

```
Private          33906
Self-emp-not-inc  3862
Local-gov        3136
?                2799
State-gov        1981
Self-emp-inc     1695
Federal-gov      1432
Without-pay      21
Never-worked     10
Name: workclass, dtype: int64
```

```
1 df['education'].value_counts()
```

```
HS-grad      15784
Some-college 10878
Bachelors    8025
Masters      2657
Assoc-voc    2061
11th         1812
Assoc-acdm   1601
10th         1389
7th-8th      955
Prof-school  834
9th          756
12th         657
Doctorate    594
5th-6th      509
1st-4th      247
Preschool    83
Name: education, dtype: int64
```

```
1 df['occupation'].value_counts()
```

```
2
```

```
Prof-specialty    6172
Craft-repair      6112
Exec-managerial   6086
Adm-clerical      5611
Sales             5504
Other-service     4923
Machine-op-inspct 3022
?                2809
Transport-moving  2355
Handlers-cleaners 2072
Farming-fishing   1490
Tech-support      1446
Protective-serv   983
Priv-house-serv   242
Armed-Forces      15
Name: occupation, dtype: int64
```

```
1 df['capital-gain'].value_counts()
```

```
2 #ตัดทิ้ง
```

```
0      44807
15024    513
7688     410
7298     364
99999    244
...
```

```

1111      1
7262      1
22040     1
1639      1
2387      1
Name: capital-gain, Length: 123, dtype: int64

```

```
1 df.groupby('income')['educational-num'].value_counts()
```

```

income  educational-num
<=50K   9                13281
        10               8815
        13               4712
        7               1720
        11              1539
        6               1302
        14              1198
        12              1188
        4                893
        5                715
        8                609
        3                482
        2                239
        15               217
        16               163
        1                82
>50K    13              3313
        9               2503
        10              2063
        14              1459
        15               617
        11               522
        16               431
        12               413
        7                92
        6                87
        4                62
        8                48
        5                41
        3                27
        2                 8
        1                 1
Name: educational-num, dtype: int64

```

```
1 df['capital-loss'].value_counts()
```

```
2 #ตัดทิ้ง
```

```

0      46560
1902     304
1977     253
1887     233
2415      72
...
2465      1
2080      1
155       1
1911      1
2201      1
Name: capital-loss, Length: 99, dtype: int64

```

```
1 df['hours-per-week'].value_counts()
```

```
2 df['hours-per-week'].describe()
```

```

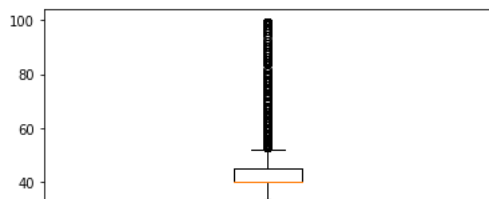
count    48842.000000
mean      40.422382
std       12.391444
min        1.000000
25%       40.000000
50%       40.000000
75%       45.000000
max       99.000000
Name: hours-per-week, dtype: float64

```

```

1 plt.boxplot(df['hours-per-week'])
2 fig = plt.figure(figsize=(10, 7))
3 plt.show()

```



```
1 q1 = np.quantile(df['hours-per-week'], 0.25)
2 q3 = np.quantile(df['hours-per-week'], 0.75)
3 print(q1,q3)
```

```
40.0 45.0
```

```
1 print(len(df[(df['hours-per-week']<=45)&(df['hours-per-week']>=40)]))
2 print(len(df[df['hours-per-week']<40]))
3 print(len(df[df['hours-per-week']>45]))
4 # แบ่งเป็นสาม class few general hard
```

```
26454
11687
10701
```

```
1 df.groupby('income')['age', 'workclass', 'fnlwgt', 'education', 'educational-num', 'marital-st
2 'capital-gain', 'capital-loss', 'hours-per-week', 'native-country'].des
```

```
<ipython-input-16-6e6c01bdb606>:1: FutureWarning: Indexing with multiple keys
df.groupby('income')['age', 'workclass', 'fnlwgt', 'education', 'educational-num', 'marital-st', 'capital-gain', 'capital-loss', 'hours-per-week', 'native-country'].des
```

	age				fnlwgt					
	count	mean	std	min	25%	50%	75%	max	count	mean
income										
<=50K	37155.0	36.872184	14.104118	17.0	25.0	34.0	46.0	90.0	37155.0	190039.5655
>50K	11687.0	44.275178	10.558983	19.0	36.0	43.0	51.0	90.0	11687.0	188470.5745

```
1 df.groupby('income')['age', 'workclass', 'fnlwgt', 'education', 'educational-num', 'marital-st
2 'capital-gain', 'capital-loss', 'hours-per-week', 'native-country'].mea
```

```
<ipython-input-21-01d05853866b>:1: FutureWarning: Indexing with multiple keys
df.groupby('income')['age', 'workclass', 'fnlwgt', 'education', 'educational-num', 'marital-st', 'capital-gain', 'capital-loss', 'hours-per-week', 'native-country'].mea
```

	age	fnlwgt	educational-num	capital-gain	capital-loss	hours-per-week
income						
<=50K	36.872184	190039.565523	9.598493	147.010308	54.151931	38.840048

ดับเบิลคลิก (หรือกด Enter) เพื่อแก้ไข

```
1 df.groupby('income')['age', 'workclass', 'fnlwgt', 'education', 'educational-num', 'marital-st
2 'capital-gain', 'capital-loss', 'hours-per-week', 'native-country'].med
```

```
<ipython-input-28-bdd548662ebf>:1: FutureWarning: Indexing with multiple keys
df.groupby('income')['age', 'workclass', 'fnlwgt', 'education', 'educational-num', 'marital-st', 'capital-gain', 'capital-loss', 'hours-per-week', 'native-country'].med
```

	age	fnlwgt	educational-num	capital-gain	capital-loss	hours-per-week
income						
<=50K	34.0	178811.0	9.0	0.0	0.0	40.0

```
1 df.head()
```

	age	workclass	fnlwgt	education	educational-num	marital-status	occupation	relat
0	25	Private	226802	11th	7	Never-married	Machine-op-inspct	
1	38	Private	89814	HS-grad	9	Married-civ-spouse	Farming-fishing	
				Assoc-		Married-	Protective-	

1 #Categories variable : race , gender

```
1 # df.loc[df['income'] == '>50K', 'age'].value_counts()
2 # df.loc[df['income'] == '>50K', 'race'].plot(kind='bar')
```

```
1 df['income_class'] = df.loc[:, 'income']
```

```
1 df
```

	age	workclass	fnlwgt	education	educational-num	marital-status	occupation	r
0	25	Private	226802	11th	7	Never-married	Machine-op-inspct	
1	38	Private	89814	HS-grad	9	Married-civ-spouse	Farming-fishing	
2	28	Local-gov	336951	Assoc-acdm	12	Married-civ-spouse	Protective-serv	
3	44	Private	160323	Some-college	10	Married-civ-spouse	Machine-op-inspct	
4	18	?	103497	Some-college	10	Never-married	?	
...	...	...	...	...	...	...	...	...
48837	27	Private	257302	Assoc-acdm	12	Married-civ-spouse	Tech-support	

```
1 #เปลี่ยนเป็น income เป็น num
2 df.loc[df['income'] == '>50K', 'income_class'] = 0
3 df.loc[df['income'] == '<=50K', 'income_class'] = 1
4 df
```

	age	workclass	fnlwgt	education	educational-num	marital-status	occupation	r
0	25	Private	226802	11th	7	Never-married	Machine-op-inspct	
1	38	Private	89814	HS-grad	9	Married-civ-spouse	Farming-fishing	
2	28	Local-gov	336951	Assoc-acdm	12	Married-civ-spouse	Protective-serv	
3	44	Private	160323	Some-college	10	Married-civ-spouse	Machine-op-inspct	
4	18	?	103497	Some-college	10	Never-married	?	
...	...	...	...	...	...	...	...	...
48837	27	Private	257302	Assoc-acdm	12	Married-civ-spouse	Tech-support	

```
1 df['relationship'].value_counts()
```

```
Husband      19716
Not-in-family 12583
Own-child     7581
Unmarried     5125
Wife          2331
```

```
Other-relative    1506
Name: relationship, dtype: int64
```

```
1 # df.groupby('income_class')['age'].plot(kind='bar')
```

```
1 # df.groupby('income').plot(kind='bar')
```

```
1 cols = ['age', 'workclass', 'fnlwgt', 'education', 'educational-num', 'marital-status', 'occup
2          'capital-gain', 'capital-loss', 'hours-per-week', 'native-country']
3 for i in cols:
4     print(df.groupby('income')[i].value_counts())
5 # age :class1 >80 , class0 <80
6 # fnlwgt :class1 >900k , class0 <900k
7 # race : White , other
8 # TODO work class, education, educational-num, marital-status, occupation
```

```
income  age
<=50K   23      1307
        24      1162
        22      1161
        25      1119
        27      1117
        ...
>50K    83         2
        20         1
        84         1
        85         1
        88         1
Name: age, Length: 142, dtype: int64
income  workclass
<=50K   Private      26519
        Self-emp-not-inc 2785
        ?             2534
        Local-gov      2209
        State-gov      1451
        Federal-gov     871
        Self-emp-inc     757
        Without-pay      19
        Never-worked     10
>50K    Private      7387
        Self-emp-not-inc 1077
        Self-emp-inc     938
        Local-gov       927
        Federal-gov     561
        State-gov       530
        ?               265
        Without-pay      2
Name: workclass, dtype: int64
income  fnlwgt
<=50K   113364      14
        190290      14
        203488      14
        117789      13
        194630      13
        ..
>50K    914061       1
        953588       1
        1033222       1
        1097453       1
        1226583       1
Name: fnlwgt, Length: 32732, dtype: int64
income  education
<=50K   HS-grad      13281
        Some-college  8815
        Bachelors     4712
        11th          1720
        Assoc-voc     1539
        10th          1302
        Masters       1198
        Assoc-acdm     1188
        7th-8th        893
        9th           715
        12th          609
        5th-6th        482
        ..
        ..
```

✓ Observe

```

1 for i in cols:
2     print(df.loc[df['income'] == '>50K',i].value_counts())
3     print(df.loc[df['income'] == '<=50K',i].value_counts())
4

```

```

46    439
47    429
41    427
39    423
37    422

```

```
...
```

```

83    2
85    1
20    1
88    1
84    1

```

```
Name: age, Length: 68, dtype: int64
```

```

23   1307
24   1162
22   1161
25   1119
27   1117

```

```
...
```

```

88    5
85    4
87    3
89    2
86    1

```

```
Name: age, Length: 74, dtype: int64
```

```

Private          7387
Self-emp-not-inc 1077
Self-emp-inc     938
Local-gov        927
Federal-gov      561
State-gov        530
?                265

```

```
Without-pay      2
```

```
Name: workclass, dtype: int64
```

```

Private          26519
Self-emp-not-inc 2785
?                2534
Local-gov        2209
State-gov        1451
Federal-gov      871
Self-emp-inc     757
Without-pay      19
Never-worked     10

```

```
Name: workclass, dtype: int64
```

```

121124    12
125892    12
148995    12
123011    12
132879    11

```

```
..
```

```

138022    1
87418     1
177307    1
270335    1
287927    1

```

```
Name: fnlwgt, Length: 8172, dtype: int64
```

```

113364    14
190290    14
203488    14

```

```
-----
```

```

1 len(df[(df['income'] == '>50K')&(df['hours-per-week'] <= 30)])
2

```

```
526
```

```

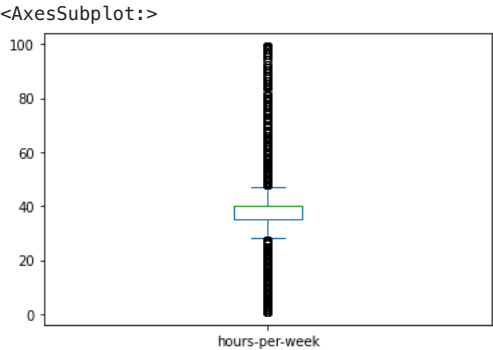
1 len(df[(df['income'] == '<=50K')&(df['hours-per-week'] < 30)])
2

```

```
5741
```

```
1 df.loc[df['income'] == '>50K', 'hours-per-week'].plot(kind='box')
```

```
<AxesSubplot:>
1 df.loc[df['income'] == '>50K', 'hours-per-week'].plot(kind='box')
2
1 df.loc[df['income'] == '<=50K', 'hours-per-week'].plot(kind='box')
2
```



```
1 df.loc[df['income'] == '<=50K', 'hours-per-week'].describe()

count    37155.000000
mean      38.840048
std       12.356849
min        1.000000
25%       35.000000
50%       40.000000
75%       40.000000
max       99.000000
Name: hours-per-week, dtype: float64
```

```
1 len(df[df['fnlwt']>900000])
18

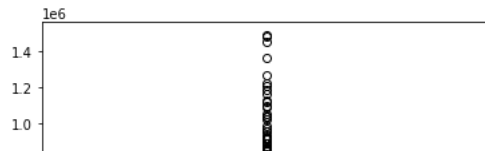
1 df[df['income']=='>50K']
2
```

	age	workclass	fnlwt	education	educational- num	marital- status	occupation	r
2	28	Local-gov	336951	Assoc- acdm	12	Married- civ-spouse	Protective- serv	
3	44	Private	160323	Some- college	10	Married- civ-spouse	Machine-op- inspct	
7	63	Self-emp- not-inc	104626	Prof-school	15	Married- civ-spouse	Prof- specialty	
10	65	Private	184454	HS-grad	9	Married- civ-spouse	Machine-op- inspct	
14	48	Private	279724	HS-grad	9	Married- civ-spouse	Machine-op- inspct	
...	...	...	...	...	...	...	...	
48820	71	?	287372	Doctorate	16	Married- civ-spouse	?	

```
1 df['fnlwt'].plot(kind='box')
```



&lt;AxesSubplot:&gt;



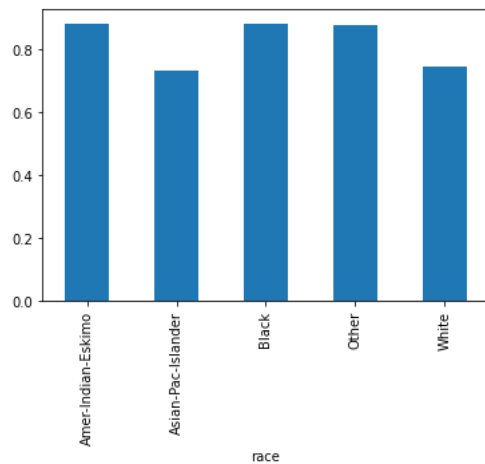
```
1 df.loc[df['income'] == '<=50K', 'capital-gain'].value_counts()
```

```
0      35611
5013     117
3325      81
2174      74
4650      63
...
1731       1
1111       1
22040      1
1639       1
2387       1
Name: capital-gain, Length: 92, dtype: int64
```

```
1 df.loc[df['income'] == '<=50K', 'capital-gain'].value_counts()
```

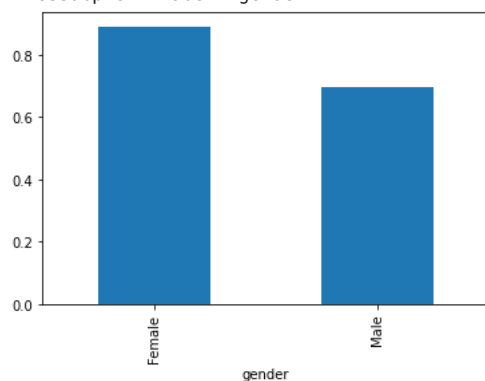
```
1 # n_col = ['race', 'gender', 'relationship']
2 df.groupby('race')['income_class'].mean().plot(kind='bar')
```

&lt;AxesSubplot:xlabel='race'&gt;



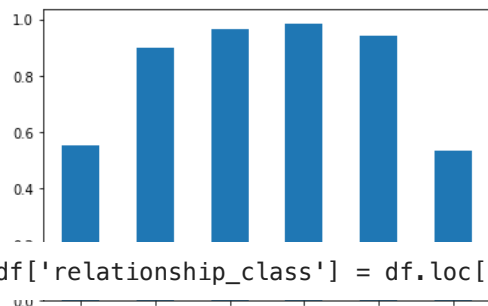
```
1 df.groupby('gender')['income_class'].mean().plot(kind='bar')
2
```

&lt;AxesSubplot:xlabel='gender'&gt;



```
1 df.groupby('relationship')['income_class'].mean().plot(kind='bar')
2 #แบ่งเป็น married(Husband,Wife) กับ other
```

&lt;AxesSubplot:xlabel='relationship'&gt;



```
1 df['relationship_class'] = df.loc[:, 'relationship']
```

```
1 df['relationship_class'] = np.where((df['relationship_class'] == 'Husband')|(df['relationship
```

```
1 df
```

	age	workclass	fnlwgt	education	educational- num	marital- status	occupation	r
0	25	Private	226802	11th	7	Never-married	Machine-op- inspct	
1	38	Private	89814	HS-grad	9	Married-civ-spouse	Farming- fishing	
2	28	Local-gov	336951	Assoc-acdm	12	Married-civ-spouse	Protective-serv	
3	44	Private	160323	Some-college	10	Married-civ-spouse	Machine-op- inspct	
4	18	?	103497	Some-college	10	Never-married	?	
...	...	...	...	...	...	...	...	...
48837	27	Private	257302	Assoc-acdm	12	Married-civ-spouse	Tech-support	

```
1 df['Adm-clerical'] = df.loc[:, 'occupation']
```

```
2 df['Adm-clerical'] = np.where(df['occupation'] == 'Adm-clerical',1,0)
```

```
3 df
```

	age	workclass	fnlwgt	education	educational- num	marital- status	occupation	r
0	25	Private	226802	11th	7	Never-married	Machine-op- inspct	
1	38	Private	89814	HS-grad	9	Married-civ-spouse	Farming- fishing	
2	28	Local-gov	336951	Assoc-acdm	12	Married-civ-spouse	Protective-serv	
3	44	Private	160323	Some-college	10	Married-civ-spouse	Machine-op- inspct	
4	18	?	103497	Some-college	10	Never-married	?	
...	...	...	...	...	...	...	...	...
48837	27	Private	257302	Assoc-acdm	12	Married-civ-spouse	Tech-support	

```
1 df['Mexico'] = df.loc[:, 'native-country']
```

```
2 df['Mexico'] = np.where(df['native-country'] == 'Mexico',1,0)
```

```
3 df
```

	age	workclass	fnlwgt	education	educational-num	marital-status	occupation	relationship	race	gender	capital-gain	capital-los
0	25	Private	226802	11th	7	Never-married	Machine-op-inspct	Own-child	Black	Male	0	
1	38	Private	89814	HS-grad	9	Married-civ-spouse	Farming-fishing	Husband	White	Male	0	
2	28	Local-gov	336951	Assoc-acdm	12	Married-civ-spouse	Protective-serv	Husband	White	Male	0	
3	44	Private	160323	Some-college	10	Married-civ-spouse	Machine-op-inspct	Husband	Black	Male	7688	
4	18	?	103497	Some-college	10	Never-married	?	Own-child	White	Female	0	
...	...	...	...	...	...	...	...	...	...	...	...	
48837	27	Private	257302	Assoc-acdm	12	Married-civ-spouse	Tech-support	Wife	White	Female	0	
48838	40	Private	154374	HS-grad	9	Married-civ-spouse	Machine-op-inspct	Husband	White	Male	0	

```

1 df['Own-child'] = df.loc[:, 'relationship']
2 df['Own-child'] = np.where(df['relationship'] == 'Own-child', 1, 0)
3 df

```

	age	workclass	fnlwgt	education	educational-num	marital-status	occupation	relationship	race	gender	capital-gain	capital-los
0	25	Private	226802	11th	7	Never-married	Machine-op-inspct	Own-child	Black	Male	0	
1	38	Private	89814	HS-grad	9	Married-civ-spouse	Farming-fishing	Husband	White	Male	0	
2	28	Local-gov	336951	Assoc-acdm	12	Married-civ-spouse	Protective-serv	Husband	White	Male	0	
3	44	Private	160323	Some-college	10	Married-civ-spouse	Machine-op-inspct	Husband	Black	Male	7688	
4	18	?	103497	Some-college	10	Never-married	?	Own-child	White	Female	0	
...	...	...	...	...	...	...	...	...	...	...	...	
48837	27	Private	257302	Assoc-acdm	12	Married-civ-spouse	Tech-support	Wife	White	Female	0	
48838	40	Private	154374	HS-grad	9	Married-civ-spouse	Machine-op-inspct	Husband	White	Male	0	
48839	58	Private	151910	HS-grad	9	Widowed	Adm-clerical	Unmarried	White	Female	0	
48840	22	Private	201490	HS-grad	9	Never-married	Adm-clerical	Own-child	White	Male	0	
...	...	Self-emp-	...	...	...	Married-	Exec-	...	...	...	...	

```

1 # utlying-US(Guam-USVI-etc)
2 df['Outlying-US(Guam-USVI-etc)'] = df.loc[:, 'native-country']
3 df['Outlying-US(Guam-USVI-etc)'] = np.where(df['native-country'] == 'Outlying-US(Guam-USVI-etc)', 1, 0)
4 df

```

	age	workclass	fnlwgt	education	educational-num	marital-status	occupation	relationship	race	gender	...	relationship
0	25	Private	226802	11th	7	Never-married	Machine-op-inspct	Own-child	Black	Male	...	
1	38	Private	89814	HS-grad	9	Married-civ-spouse	Farming-fishing	Husband	White	Male	...	
2	28	Local-gov	336951	Assoc-acdm	12	Married-civ-spouse	Protective-serv	Husband	White	Male	...	
3	44	Private	160323	Some-college	10	Married-civ-spouse	Machine-op-inspct	Husband	Black	Male	...	

```
1 df['work_hard'] = df.loc[:, 'hours-per-week']
2 df['work_hard'] = np.where(df['hours-per-week'] < 30, 1, 0)
3 df
```

	age	workclass	fnlwgt	education	educational-num	marital-status	occupation	relationship	race	gender	...	relationship
0	25	Private	226802	11th	7	Never-married	Machine-op-inspct	Own-child	Black	Male	...	
1	38	Private	89814	HS-grad	9	Married-civ-spouse	Farming-fishing	Husband	White	Male	...	
2	28	Local-gov	336951	Assoc-acdm	12	Married-civ-spouse	Protective-serv	Husband	White	Male	...	
3	44	Private	160323	Some-college	10	Married-civ-spouse	Machine-op-inspct	Husband	Black	Male	...	
4	18	?	103497	Some-college	10	Never-married	?	Own-child	White	Female	...	
...	...	...	...	...	...	...	...	...	...	...	...	
48837	27	Private	257302	Assoc-acdm	12	Married-civ-spouse	Tech-support	Wife	White	Female	...	
48838	40	Private	154374	HS-grad	9	Married-civ-spouse	Machine-op-inspct	Husband	White	Male	...	
48839	58	Private	151910	HS-grad	9	Widowed	Adm-clerical	Unmarried	White	Female	...	
48840	22	Private	201490	HS-grad	9	Never-married	Adm-clerical	Own-child	White	Male	...	
48841	52	Self-emp-inc	287927	HS-grad	9	Married-civ-spouse	Exec-managerial	Wife	White	Female	...	

```
1 df['capg_h'] = df.loc[:, 'capital-gain']
2 df['capg_h'] = np.where(df['capital-gain'] > 5100, 1, 0)
3 df
```

	age	workclass	fnlwgt	education	educational-num	marital-status	occupation	r
0	25	Private	226802	11th	7	Never-married	Machine-op-inspct	
1	38	Private	89814	HS-grad	9	Married-civ-spouse	Farming-fishing	
2	28	Local-gov	336951	Assoc-acdm	12	Married-civ-spouse	Protective-serv	
3	44	Private	160323	Some-college	10	Married-civ-spouse	Machine-op-inspct	
4	18	?	103497	Some-college	10	Never-married	?	

```

1 df['Never-married'] = df.loc[:, 'marital-status']
2 df['Never-married'] = np.where(df['marital-status'] == 'Never-married', 1, 0)
3 df

```

	age	workclass	fnlwgt	education	educational-num	marital-status	occupation	r
0	25	Private	226802	11th	7	Never-married	Machine-op-inspct	
1	38	Private	89814	HS-grad	9	Married-civ-spouse	Farming-fishing	
2	28	Local-gov	336951	Assoc-acdm	12	Married-civ-spouse	Protective-serv	
3	44	Private	160323	Some-college	10	Married-civ-spouse	Machine-op-inspct	
4	18	?	103497	Some-college	10	Never-married	?	
...	...	...	...	...	...	...	...	...
48837	27	Private	257302	Assoc-acdm	12	Married-civ-spouse	Tech-support	
48838	40	Private	154374	HS-grad	9	Married-civ-spouse	Machine-op-inspct	
48839	58	Private	151910	HS-grad	9	Widowed	Adm-clerical	
48840	22	Private	201490	HS-grad	9	Never-married	Adm-clerical	
48841	52	Self-emp-inc	287927	HS-grad	9	Married-civ-spouse	Exec-managerial	

```

1 df['pre_school'] = df.loc[:, 'education']
2 df['pre_school'] = np.where(df['education'] == 'Preschool', 1, 0)
3 df

```

	age	workclass	fnlwgt	education	educational-num	marital-status	occupation	r
0	25	Private	226802	11th	7	Never-married	Machine-op-inspct	
1	38	Private	89814	HS-grad	9	Married-civ-spouse	Farming-fishing	
2	28	Local-gov	336951	Assoc-acdm	12	Married-civ-spouse	Protective-serv	
3	44	Private	160323	Some-college	10	Married-civ-spouse	Machine-op-inspct	
4	18	?	103497	Some-college	10	Never-married	?	
...	...	...	...	...	...	...	...	...
48837	27	Private	257302	Assoc-acdm	12	Married-civ-spouse	Tech-support	
48838	40	Private	154374	HS-grad	9	Married-civ-spouse	Machine-op-inspct	
48839	58	Private	151910	HS-grad	9	Widowed	Adm-clerical	
48840	22	Private	201490	HS-grad	9	Never-married	Adm-clerical	
48841	52	Self-emp-inc	287927	HS-grad	9	Married-civ-spouse	Exec-managerial	

```
1 df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 48842 entries, 0 to 48841
Data columns (total 26 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   age                                   48842 non-null  int64
1   workclass                            48842 non-null  object
2   fnlwgt                               48842 non-null  int64
3   education                            48842 non-null  object
4   educational-num                       48842 non-null  int64
5   marital-status                       48842 non-null  object
6   occupation                           48842 non-null  object
7   relationship                         48842 non-null  object
8   race                                 48842 non-null  object
9   gender                               48842 non-null  object
10  capital-gain                         48842 non-null  int64
11  capital-loss                         48842 non-null  int64
12  hours-per-week                       48842 non-null  int64
13  native-country                      48842 non-null  object
14  income                               48842 non-null  object
15  income_class                         48842 non-null  int64
16  relationship_class                   48842 non-null  object
17  Adm-clerical                        48842 non-null  int64
18  Mexico                              48842 non-null  int64
19  Own-child                           48842 non-null  int64
20  utlying-US(Guam-USVI-etc)          48842 non-null  int64
21  work_hard                           48842 non-null  int64
22  capg_h                              48842 non-null  int64
23  Never-married                       48842 non-null  int64
24  pre_school                          48842 non-null  int64
25  Outlying-US(Guam-USVI-etc)         48842 non-null  int64
dtypes: int64(16), object(10)
memory usage: 9.7+ MB
```

```
1 df['income_class'] = df['income_class'].astype(int)
```

```
1 df.corr()
```

	age	fnlwgt	educational-num	capital-gain	capital-loss	hours-per-week	inc
age	1.000000	-0.076628	0.030940	0.077229	0.056944	0.071558	
fnlwgt	-0.076628	1.000000	-0.038761	-0.003706	-0.004366	-0.013519	
educational-num	0.030940	-0.038761	1.000000	0.125146	0.080972	0.143689	
capital-gain	0.077229	-0.003706	0.125146	1.000000	-0.031441	0.082157	
capital-loss	0.056944	-0.004366	0.080972	-0.031441	1.000000	0.054467	
hours-per-week	0.071558	-0.013519	0.143689	0.082157	0.054467	1.000000	
income_class	-0.230369	0.006339	-0.332613	-0.223013	-0.147554	-0.227687	
Adm-clerical	-0.038116	0.007480	0.004142	-0.029105	-0.021457	-0.078916	
Mexico	-0.051478	0.126589	-0.222085	-0.012540	-0.019178	-0.002376	
Own-child	-0.432990	0.016716	-0.097316	-0.052038	-0.049167	-0.251827	
utlying-US(Guam-USVI-etc)	NaN	NaN	NaN	NaN	NaN	NaN	
work_hard	-0.053907	-0.010005	-0.091675	-0.038148	-0.034280	-0.692304	
capg_h	0.115607	-0.003924	0.162125	0.576914	-0.048619	0.098459	
.....							

Test 1

Logistic Regression model

```
1 sex = pd.get_dummies(df['gender'])
```

```
1 sex.head(3)
```

	Female	Male
0	0	1
1	0	1
2	0	1

```
1 df_new = df.drop(['gender'],axis=1)
```

```
1 df_new = pd.concat([df_new,sex],axis=1)
```

```
1 df_new.head()
```

	age	workclass	fnlwgt	education	educational-num	marital-status	occupation	relat
0	25	Private	226802	11th	7	Never-married	Machine-op-inspct	
1	38	Private	89814	HS-grad	9	Married-civ-spouse	Farming-fishing	
2	28	Local-gov	336951	Assoc-acdm	12	Married-civ-spouse	Protective-serv	
3	44	Private	160323	Some-college	10	Married-civ-spouse	Machine-op-inspct	
4	18	?	103497	Some-college	10	Never-married	?	

```
1 df_new.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 48842 entries, 0 to 48841
Data columns (total 27 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   age                                   48842 non-null  int64
1   workclass                             48842 non-null  object
2   fnlwgt                                48842 non-null  int64
3   education                             48842 non-null  object
4   educational-num                        48842 non-null  int64
5   marital-status                        48842 non-null  object
6   occupation                             48842 non-null  object
7   relationship                           48842 non-null  object
8   race                                   48842 non-null  object
9   capital-gain                           48842 non-null  int64
10  capital-loss                           48842 non-null  int64
11  hours-per-week                         48842 non-null  int64
12  native-country                        48842 non-null  object
13  income                                 48842 non-null  object
14  income_class                           48842 non-null  int64
15  relationship_class                     48842 non-null  object
16  Adm-clerical                           48842 non-null  int64
17  Mexico                                 48842 non-null  int64
18  Own-child                              48842 non-null  int64
19  utlying-US(Guam-USVI-etc)             48842 non-null  int64
20  work_hard                              48842 non-null  int64
21  capg_h                                 48842 non-null  int64
22  Never-married                          48842 non-null  int64
23  pre_school                             48842 non-null  int64
24  Outlying-US(Guam-USVI-etc)            48842 non-null  int64
25  Female                                 48842 non-null  uint8
26  Male                                   48842 non-null  uint8
dtypes: int64(16), object(9), uint8(2)
memory usage: 9.4+ MB
```

```
1 race = pd.get_dummies(df['race'])
```

```
1 df_new = df_new.drop(['race'],axis=1)
2 df_new = pd.concat([df_new,race],axis=1)
3 df_new.head()
```

	age	workclass	fnlwgt	education	educational-num	marital-status	occupation	relat
0	25	Private	226802	11th	7	Never-married	Machine-op-inspct	
1	38	Private	89814	HS-grad	9	Married-civ-spouse	Farming-fishing	
2	28	Local-gov	336951	Assoc-acdm	12	Married-civ-spouse	Protective-serv	
3	44	Private	160323	Some-college	10	Married-civ-spouse	Machine-op-inspct	
4	18	?	103497	Some-college	10	Never-married	?	

```
1 rela = pd.get_dummies(df['relationship_class'])
```

```
1 df_new = df_new.drop(['relationship_class'],axis=1)
2 df_new = pd.concat([df_new,rela],axis=1)
3 df_new.head()
```



	age	workclass	fnlwgt	education	educational-num	marital-status	occupation	relat
0	25	Private	226802	11th	7	Never-married	Machine-op-inspct	
1	38	Private	89814	HS-grad	9	Married-civ-spouse	Farming-fishing	
2	28	Local-gov	336951	Assoc-acdm	12	Married-civ-spouse	Protective-serv	
3	44	Private	160323	Some-college	10	Married-civ-spouse	Machine-op-inspct	
4	18	?	103497	Some-college	10	Never-married	?	

## ✓ Train

```
1 df.corr()
```

	age	fnlwgt	educational-num	capital-gain	capital-loss	hours-per-week	inc
age	1.000000	-0.076628	0.030940	0.077229	0.056944	0.071558	
fnlwgt	-0.076628	1.000000	-0.038761	-0.003706	-0.004366	-0.013519	
educational-num	0.030940	-0.038761	1.000000	0.125146	0.080972	0.143689	
capital-gain	0.077229	-0.003706	0.125146	1.000000	-0.031441	0.082157	
capital-loss	0.056944	-0.004366	0.080972	-0.031441	1.000000	0.054467	
hours-per-week	0.071558	-0.013519	0.143689	0.082157	0.054467	1.000000	
income_class	-0.230369	0.006339	-0.332613	-0.223013	-0.147554	-0.227687	
Adm-clerical	-0.038116	0.007480	0.004142	-0.029105	-0.021457	-0.078916	
Mexico	-0.051478	0.126589	-0.222085	-0.012540	-0.019178	-0.002376	
Own-child	-0.432990	0.016716	-0.097316	-0.052038	-0.049167	-0.251827	
utlying-US(Guam-USVI-etc)	NaN	NaN	NaN	NaN	NaN	NaN	
work_hard	-0.053907	-0.010005	-0.091675	-0.038148	-0.034280	-0.692304	
capg_h	0.115607	-0.003924	0.162125	0.576914	-0.048619	0.098459	
Never-							

```
1 cor = df.corr()
2 cor_target = abs(cor['income_class'])
3
4 #Selecting highly correlated features
5 relevant_features = cor_target[cor_target>0.2]
6 relevant_features
```

```
age          0.230369
educational-num 0.332613
capital-gain  0.223013
```

```

hours-per-week    0.227687
income_class      1.000000
Own-child         0.225691
capg_h            0.371346
Never-married     0.318782
Name: income_class, dtype: float64

```

```
1 from sklearn.model_selection import train_test_split
```

```

1 # col_pre = ['Female', 'Male', 'Black', 'White']
2 # col_pre = ['capital-gain', 'capital-loss', 'Adm-clerical', 'Mexico', 'Own-child',]
3 # col_pre = ['educational-num', 'capg_h', 'age', 'Never-married']
4 col_pre = ['educational-num', 'capg_h', 'Never-married']
5
6 # col_pre = ['capital-gain', 'capital-loss', 'Never-married']
7

```

```

1 X = df_new[col_pre]
2 y = df_new['income_class']
3 y = y.astype('int')
4 X_train, X_test, y_train, y_test = train_test_split(X, y, stratify=y, test_size=0.30, random_

```

```
1 from sklearn.linear_model import LogisticRegression
```

```

1 model = LogisticRegression()
2 model.fit(X_train, y_train)

```

```
LogisticRegression()
```

```
1 y_pred = model.predict(X_test)
```

```

1 print(list(y_test[:5]))
2 print(y_pred[:5])

```

```

[1, 1, 1, 1, 1]
[1 1 1 1 0]

```

## ✓ Evaluate Model

```

1 from sklearn.metrics import confusion_matrix
2 from sklearn.metrics import classification_report

```

```
1 confusion_matrix(y_test, y_pred)
```

```

array([[ 1908,  1598],
       [ 1000, 10147]])

```

```
1 print(classification_report(y_test, y_pred))
```

	precision	recall	f1-score	support
0	0.66	0.54	0.59	3506
1	0.86	0.91	0.89	11147
accuracy			0.82	14653
macro avg	0.76	0.73	0.74	14653
weighted avg	0.81	0.82	0.82	14653

## ✓ Compute odd ratio

```

1 print(X.columns)
2 print(model.intercept_)
3 print(model.coef_)

```

```

Index(['educational-num', 'capg_h', 'Never-married'], dtype='object')
[4.75837117]
[[-0.36937834 -4.14552532  2.57254605]]

```

