

```
import pandas as pd
column_names = ["age", "workclass", "fnlwgt", "education", "education-
num", "marital-status", "occupation", "relationship", "race", "sex",
"capital-gain", "capital-loss", "hours-per-week", "native-country",
"income"]
data = pd.read_csv('adult.csv', names = column_names)
data
```

	age	workclass	fnlwgt	education	education-num \
0	39	State-gov	77516	Bachelors	13
1	50	Self-emp-not-inc	83311	Bachelors	13
2	38	Private	215646	HS-grad	9
3	53	Private	234721	11th	7
4	28	Private	338409	Bachelors	13
...
32556	27	Private	257302	Assoc-acdm	12
32557	40	Private	154374	HS-grad	9
32558	58	Private	151910	HS-grad	9
32559	22	Private	201490	HS-grad	9
32560	52	Self-emp-inc	287927	HS-grad	9

	marital-status	occupation	relationship	race
0	Never-married	Adm-clerical	Not-in-family	White
1	Married-civ-spouse	Exec-managerial	Husband	White
2	Divorced	Handlers-cleaners	Not-in-family	White
3	Married-civ-spouse	Handlers-cleaners	Husband	Black
4	Married-civ-spouse	Prof-specialty	Wife	Black
...
32556	Married-civ-spouse	Tech-support	Wife	White
32557	Married-civ-spouse	Machine-op-inspct	Husband	White
32558	Widowed	Adm-clerical	Unmarried	White
32559	Never-married	Adm-clerical	Own-child	White
32560	Married-civ-spouse	Exec-managerial	Wife	White

	sex	capital-gain	capital-loss	hours-per-week	native-country \
0	Male	2174	0	40	United-States
1	Male	0	0	13	United-

States					
2	Male	0	0	40	United-
States					
3	Male	0	0	40	United-
States					
4	Female	0	0	40	
Cuba					
...	
...					
32556	Female	0	0	38	United-
States					
32557	Male	0	0	40	United-
States					
32558	Female	0	0	40	United-
States					
32559	Male	0	0	20	United-
States					
32560	Female	15024	0	40	United-
States					

	income
0	<=50K
1	<=50K
2	<=50K
3	<=50K
4	<=50K
...	...
32556	<=50K
32557	>50K
32558	<=50K
32559	<=50K
32560	>50K

[32561 rows x 15 columns]

data

	age	workclass	fnlwgt	education	education-num \
0	39	State-gov	77516	Bachelors	13
1	50	Self-emp-not-inc	83311	Bachelors	13
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country	sex	capital-gain	capital-loss	hours-per-week	native-	
\	0	Male	2174	0	40	United-
	1	Male	0	0	13	United-
	2	Male	0	0	40	United-
	3	Male	0	0	40	United-
	4	Female	0	0	40	
Cuba						
...		
...						
	32556	Female	0	0	38	United-
	32557	Male	0	0	40	United-
	32558	Female	0	0	40	United-
	32559	Male	0	0	20	United-
	32560	Female	15024	0	40	United-
States						

```

income
0      <=50K
1      <=50K
2      <=50K
3      <=50K
4      <=50K
...
32556  <=50K
32557  >50K
32558  <=50K
32559  <=50K
32560  >50K

```

```
[32561 rows x 15 columns]
```

#What is the central tendency of the age distribution?

```
data.describe()
```

```

              age      fnlwgt  education-num  capital-gain
capital-loss \
count  32561.000000  3.256100e+04   32561.000000  32561.000000
32561.000000
mean      38.581647  1.897784e+05      10.080679   1077.648844
87.303830
std       13.640433  1.055500e+05      2.572720   7385.292085
402.960219
min       17.000000  1.228500e+04      1.000000      0.000000
0.000000
25%       28.000000  1.178270e+05      9.000000      0.000000
0.000000
50%       37.000000  1.783560e+05     10.000000      0.000000
0.000000
75%       48.000000  2.370510e+05     12.000000      0.000000
0.000000
max       90.000000  1.484705e+06     16.000000  99999.000000
4356.000000

```

```

hours-per-week
count      32561.000000
mean        40.437456
std         12.347429
min          1.000000
25%         40.000000
50%         40.000000
75%         45.000000
max         99.000000

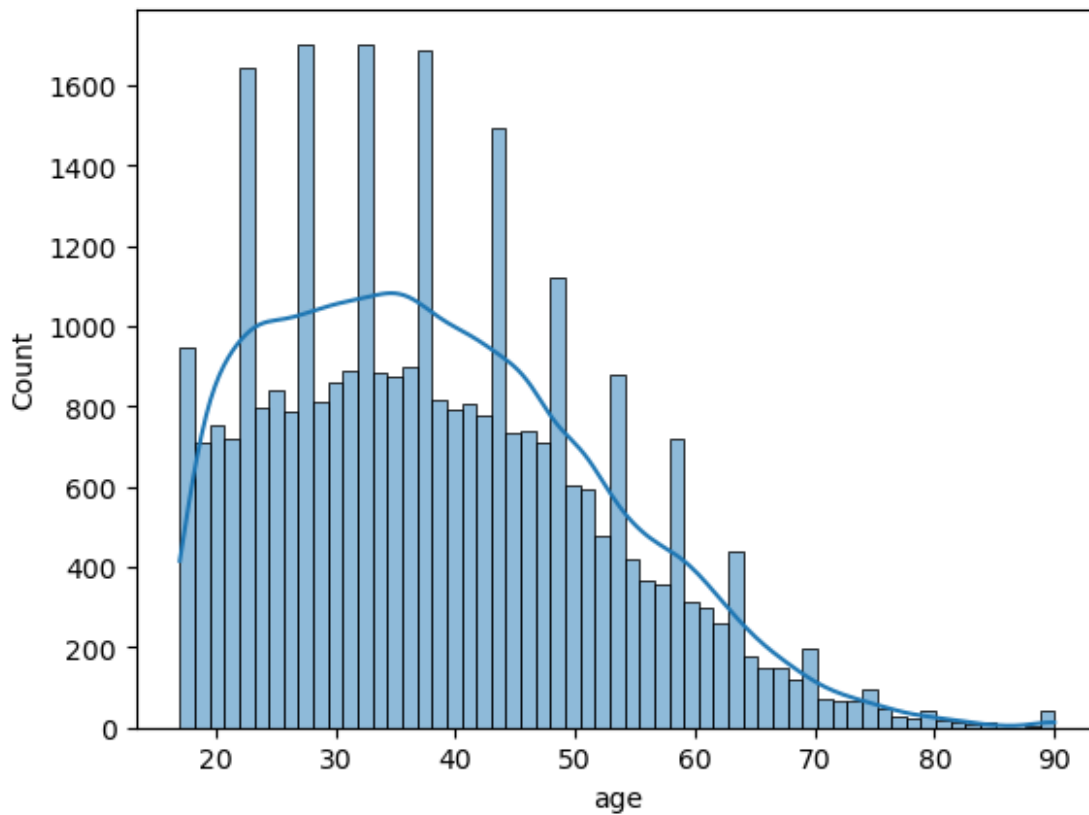
```

```

import seaborn as sns
sns.histplot(data, x='age', kde=True)

```

<Axes: xlabel='age', ylabel='Count'>



#Is the age distribution skewed or symmetric?
#Ans: Thus Age Distribution is Right Skewed
#Above

```
age_mean = data['age'].mean()  
age_mean
```

```
38.58164675532078
```

```
data.dtypes
```

age	int64
workclass	object
fnlwgt	int64
education	object
education-num	int64
marital-status	object
occupation	object
relationship	object
race	object
sex	object
capital-gain	int64

```
capital-loss      int64
hours-per-week    int64
native-country    object
income            object
dtype: object
```

#Outliers

```
q1 = data['age'].quantile(0.25)
q3 = data['age'].quantile(0.75)
IQR = q3 - q1
upper_limit = q3 + 1.5*IQR
lower_limit = q1 - 1.5*IQR
out_data = data.loc[(data['age'] > lower_limit) & (data['age'] <
upper_limit)]
out_data
```

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32560	Married-civ-spouse	Exec-managerial	Wife	White
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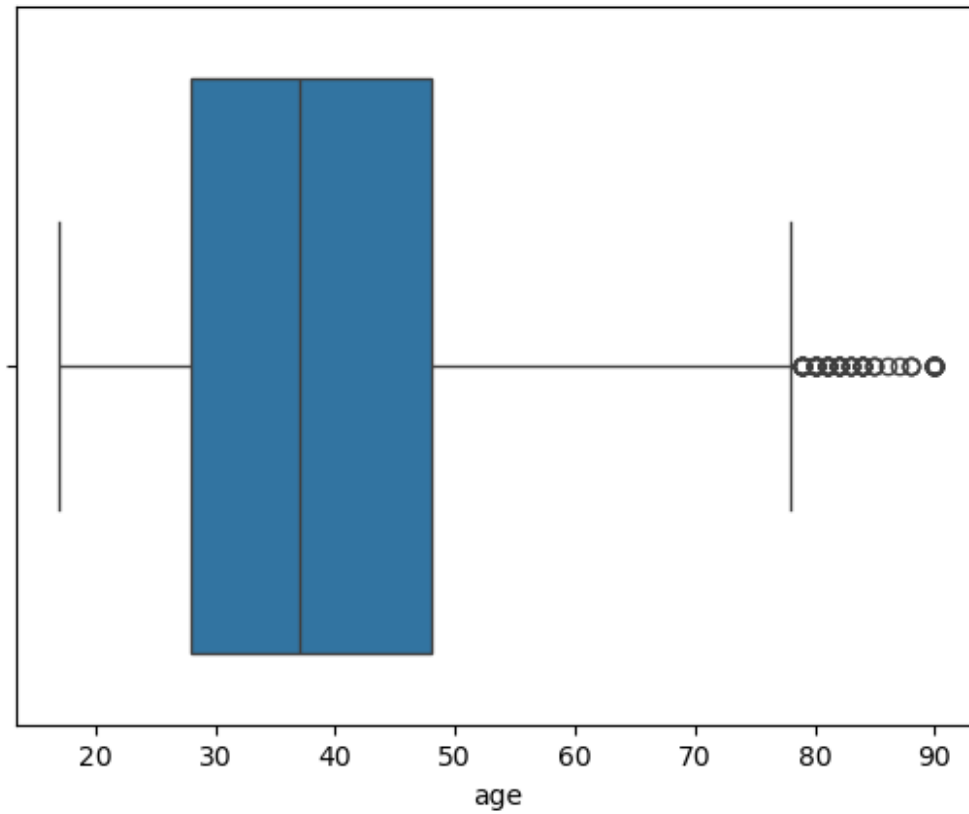
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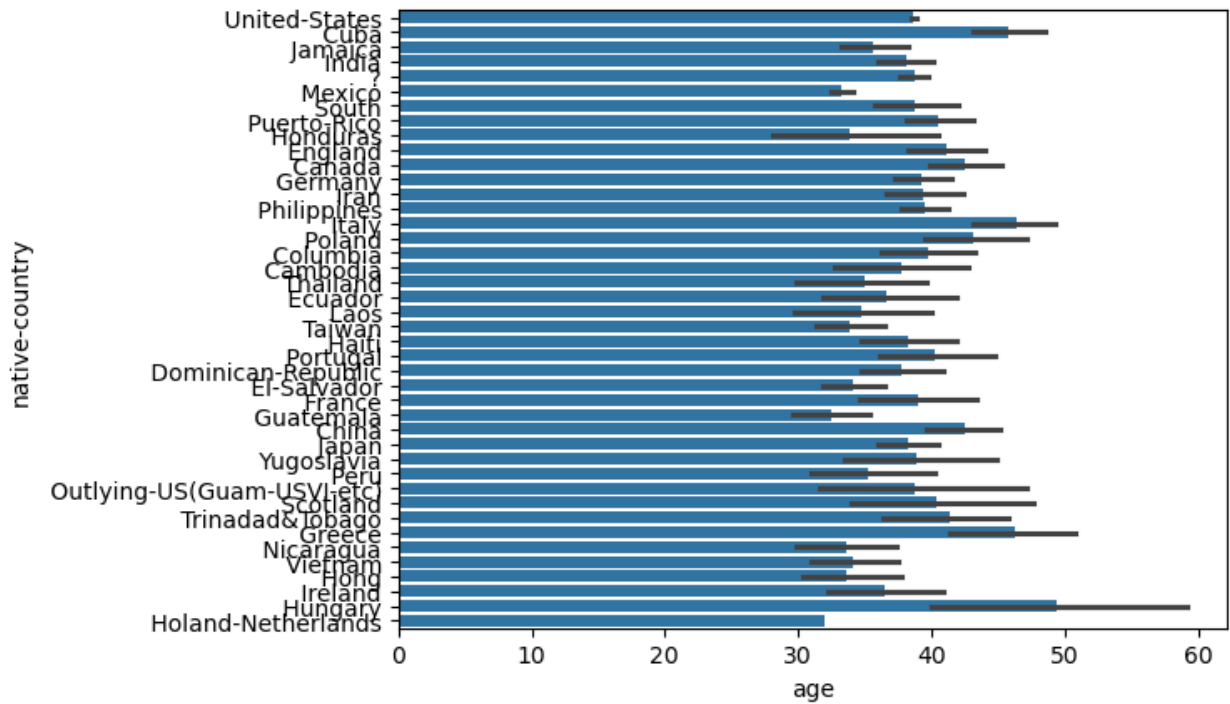
[32395 rows x 15 columns]

```
sns.boxplot(data=data,x='age')
```

```
<Axes: xlabel='age'>
```



```
sns.barplot(data,x='age',y='native-country')  
<Axes: xlabel='age', ylabel='native-country'>
```

```
sns.barplot(data,x='age',y='workclass')
<Axes: xlabel='age', ylabel='workclass'>
```

