In [1]: import pandas as pd

In [2]: data=pd.read\_csv(r"C:\Users\lavan\Downloads\adult 3.csv")

In [3]: data

Out[3]:

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

48842 rows × 15 columns

In [4]: data.shape

Out[4]: (48842, 15)

In [5]: data.head(7)

## Out[5]:

	age	workclass	fnlwgt	education	educational- num	marital- status	occupation	relationship	race	ge
0	25	Private	226802	11th	7	Never- married	Machine- op-inspct	Own-child	Black	
1	38	Private	89814	HS-grad	9	Married- civ- spouse	Farming- fishing	Husband	White	
2	28	Local-gov	336951	Assoc- acdm	12	Married- civ- spouse	Protective- serv	Husband	White	
3	44	Private	160323	Some- college	10	Married- civ- spouse	Machine- op-inspct	Husband	Black	
4	18	?	103497	Some- college	10	Never- married	?	Own-child	White	Fe
5	34	Private	198693	10th	6	Never- married	Other- service	Not-in-family	White	
6	29	?	227026	HS-grad	9	Never- married	?	Unmarried	Black	
4										

In [6]: data.tail(7)

## Out[6]:

	age	workclass	fnlwgt	education	educational- num	marital- status	occupation	relationship	rac
48835	53	Private	321865	Masters	14	Married- civ- spouse	Exec- managerial	Husband	Whit
48836	22	Private	310152	Some- college	10	Never- married	Protective- serv	Not-in-family	Whit
48837	27	Private	257302	Assoc- acdm	12	Married- civ- spouse	Tech- support	Wife	Whit
48838	40	Private	154374	HS-grad	9	Married- civ- spouse	Machine- op-inspct	Husband	Whit
48839	58	Private	151910	HS-grad	9	Widowed	Adm- clerical	Unmarried	Whit
48840	22	Private	201490	HS-grad	9	Never- married	Adm- clerical	Own-child	Whit
48841	52	Self-emp- inc	287927	HS-grad	9	Married- civ- spouse	Exec- managerial	Wife	Whit

In [7]: data.isna()

Out[7]:

	age	workclass	fnlwgt	education	educational- num	marital- status	occupation	relationship	race
0	False	False	False	False	False	False	False	False	False
1	False	False	False	False	False	False	False	False	False
2	False	False	False	False	False	False	False	False	False
3	False	False	False	False	False	False	False	False	False
4	False	False	False	False	False	False	False	False	False
48837	False	False	False	False	False	False	False	False	False
48838	False	False	False	False	False	False	False	False	False
48839	False	False	False	False	False	False	False	False	False
48840	False	False	False	False	False	False	False	False	False
48841	False	False	False	False	False	False	False	False	False

48842 rows × 15 columns

In [8]: data.isna().sum()

Out[8]: age

0 workclass 0 fnlwgt 0 education 0 educational-num 0 marital-status 0 occupation 0 relationship 0 0 race 0 gender capital-gain 0 capital-loss 0 hours-per-week 0 native-country 0 income 0 dtype: int64

```
In [9]: print(data.occupation.value_counts())
```

occupation 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: count, dtype: int64

## In [10]: print(data.gender.value\_counts())

gender

Male 32650 Female 16192

Name: count, dtype: int64

## In [11]: | print(data['marital-status'].value\_counts())

marital-status Married-civ-spouse 22379 Never-married 16117 Divorced 6633 Separated 1530 Widowed 1518 Married-spouse-absent 628 Married-AF-spouse 37 Name: count, dtype: int64

```
In [12]:
         print(data['education'].value_counts())
         education
         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: count, dtype: int64
In [13]: |print(data['workclass'].value_counts())
         workclass
         Private
                              33906
         Self-emp-not-inc
                               3862
         Local-gov
                               3136
                               2799
         State-gov
                               1981
         Self-emp-inc
                               1695
         Federal-gov
                               1432
         Without-pay
                                 21
                                 10
         Never-worked
         Name: count, dtype: int64
         data.occupation.replace({'?':'Others'},inplace=True)
In [14]:
In [15]:
         print(data.occupation.value_counts())
         occupation
         Prof-specialty
                               6172
         Craft-repair
                               6112
         Exec-managerial
                               6086
         Adm-clerical
                               5611
         Sales
                               5504
         Other-service
                               4923
         Machine-op-inspct
                               3022
         Others
                               2809
         Transport-moving
                               2355
         Handlers-cleaners
                               2072
         Farming-fishing
                               1490
         Tech-support
                               1446
         Protective-serv
                                983
         Priv-house-serv
                                242
         Armed-Forces
                                 15
         Name: count, dtype: int64
```

In [16]: data

Out[16]:

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

48842 rows × 15 columns

In [17]: data.workclass.replace({'?':'NotListed'},inplace=True)

In [18]: print(data['workclass'].value\_counts())

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

```
In [19]:
         data=data[data['workclass']!='without-pay']
         data=data[data['workclass']!='Never-worked']
In [20]: print(data['workclass'].value_counts())
         workclass
         Private
                              33906
         Self-emp-not-inc
                               3862
         Local-gov
                               3136
         NotListed
                               2799
         State-gov
                               1981
         Self-emp-inc
                               1695
                               1432
         Federal-gov
         Without-pay
                                 21
         Name: count, dtype: int64
In [21]: | data.shape
Out[21]: (48832, 15)
In [22]: | data=data[data['education']!='5th-6th']
         data=data[data['education']!='1st-4th']
         data=data[data['education']!='preschool']
In [23]: print(data['educational-num'].value_counts())
         educational-num
         9
                15782
         10
                10876
         13
                8025
         14
                 2657
         11
                 2061
         7
                 1809
         12
                 1601
         6
                 1387
         4
                  954
         15
                  834
         5
                  756
         8
                  657
         16
                  594
         1
                   83
         Name: count, dtype: int64
In [24]: data.shape
Out[24]: (48076, 15)
In [25]: data.drop(columns=['education'],inplace=True)
```

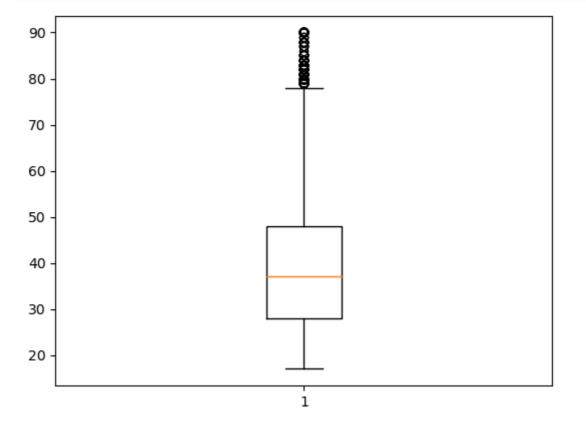
In [26]: data

Out[26]:

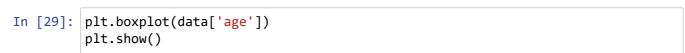
	age	workclass	fnlwgt	educational- num	marital- status	occupation	relationship	race	gender
0	25	Private	226802	7	Never- married	Machine- op-inspct	Own-child	Black	Male
1	38	Private	89814	9	Married- civ- spouse	Farming- fishing	Husband	White	Male
2	28	Local-gov	336951	12	Married- civ- spouse	Protective-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

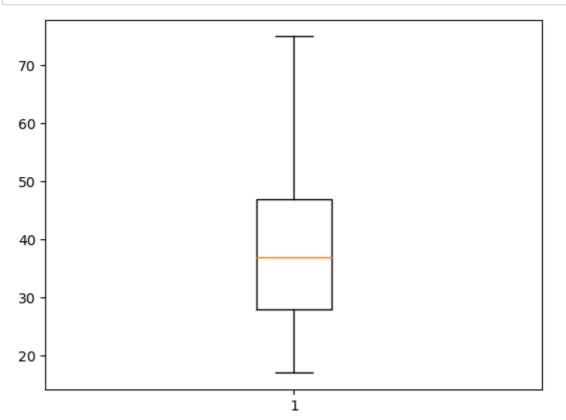
48076 rows × 14 columns

```
In [27]: import matplotlib.pyplot as plt
plt.boxplot(data['age'])
plt.show()
```



```
In [28]: data= data[(data['age']<=75) & (data['age']>=17)]
```





60000

40000

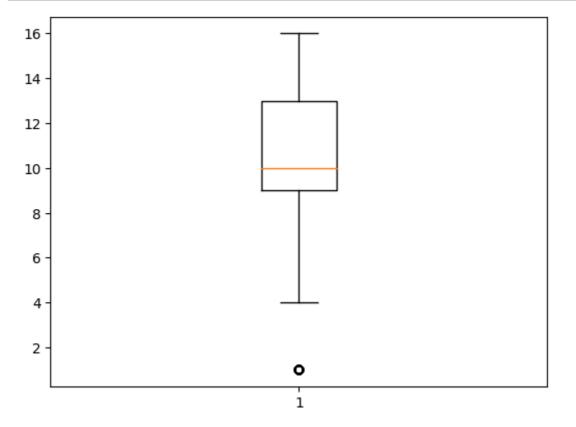
20000

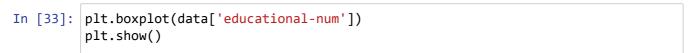
0

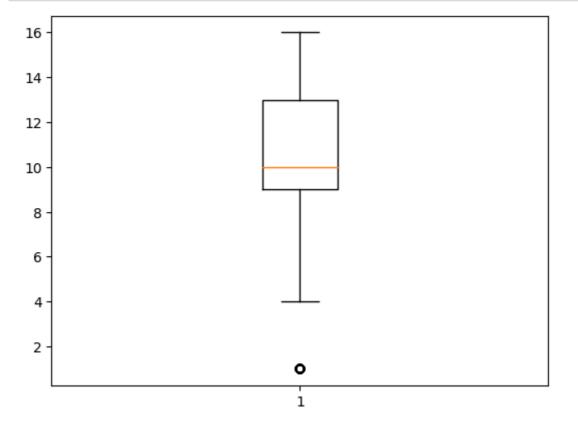
0

1

```
In [32]: plt.boxplot(data['educational-num'])
    plt.show()
```

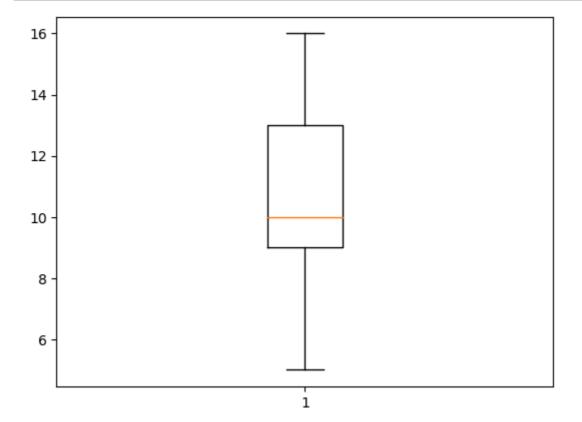


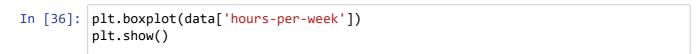


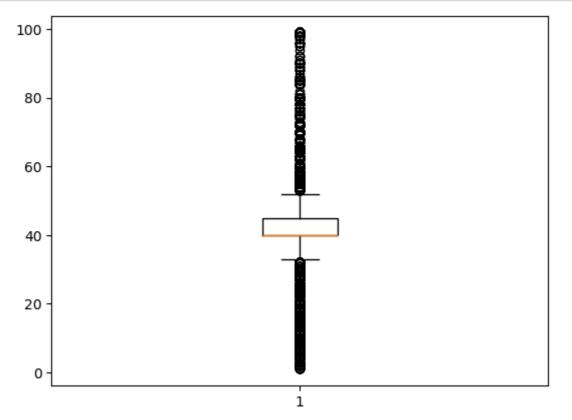


```
In [34]: data=data[(data['educational-num']<=16)&(data['educational-num']>=5)]
```

```
In [35]: plt.boxplot(data['educational-num'])
   plt.show()
```







In [37]: data.shape

Out[37]: (46739, 14)

In [38]: data

Out[38]:

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

46739 rows × 14 columns

```
In [39]: from sklearn.preprocessing import LabelEncoder
encoder=LabelEncoder()
    data['workclass']=encoder.fit_transform(data['workclass'])
    data['marital-status']=encoder.fit_transform(data['marital-status'])
    data['occupation']=encoder.fit_transform(data['occupation'])
    data['relationship']=encoder.fit_transform(data['relationship'])
    data['race']=encoder.fit_transform(data['race'])
    data['gender']=encoder.fit_transform(data['gender'])
    data['native-country']=encoder.fit_transform(data['native-country'])
```

In [40]: data

Out[40]:

	age	workclass	fnlwgt	educational- num	marital- status	occupation	relationship	race	gender	Ci
0	25	3	226802	7	4	6	3	2	1	
1	38	3	89814	9	2	4	0	4	1	
2	28	1	336951	12	2	11	0	4	1	
3	44	3	160323	10	2	6	0	2	1	
4	18	2	103497	10	4	8	3	4	0	
							•••			
48837	27	3	257302	12	2	13	5	4	0	
48838	40	3	154374	9	2	6	0	4	1	
48839	58	3	151910	9	6	0	4	4	0	
48840	22	3	201490	9	4	0	3	4	1	
48841	52	4	287927	9	2	3	5	4	0	

46739 rows × 14 columns

In [41]: x=data.drop(columns=['income'])
y=data['income']

In [42]: x

Out[42]:

	age	workclass	fnlwgt	educational- num	marital- status	occupation	relationship	race	gender	Ci
0	25	3	226802	7	4	6	3	2	1	
1	38	3	89814	9	2	4	0	4	1	
2	28	1	336951	12	2	11	0	4	1	
3	44	3	160323	10	2	6	0	2	1	
4	18	2	103497	10	4	8	3	4	0	
48837	27	3	257302	12	2	13	5	4	0	
48838	40	3	154374	9	2	6	0	4	1	
48839	58	3	151910	9	6	0	4	4	0	
48840	22	3	201490	9	4	0	3	4	1	
48841	52	4	287927	9	2	3	5	4	0	

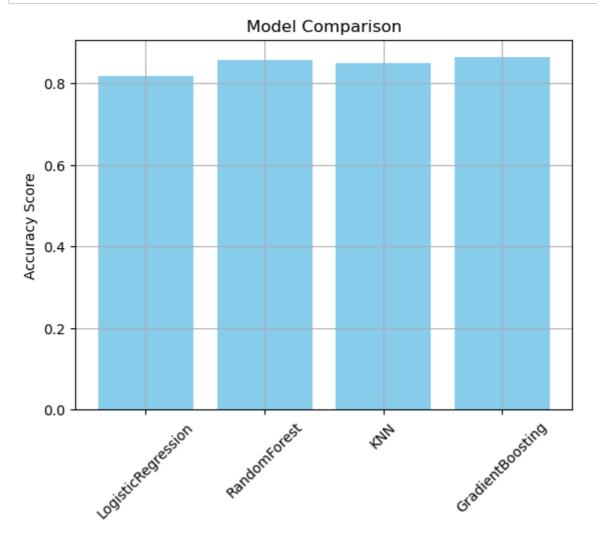
46739 rows × 13 columns

```
In [43]:
         from sklearn.pipeline import Pipeline
         from sklearn.model_selection import train_test_split
         from sklearn.metrics import accuracy_score, classification_report
         from sklearn.linear_model import LogisticRegression
         from sklearn.ensemble import RandomForestClassifier, GradientBoostingClassifie
         from sklearn.neighbors import KNeighborsClassifier
         from sklearn.svm import SVC
         from sklearn.preprocessing import StandardScaler, OneHotEncoder
         X_train, X_test, y_train, y_test = train_test_split(x, y, test_size=0.2, rando
         models = {
             "LogisticRegression": LogisticRegression(),
             "RandomForest": RandomForestClassifier(),
             "KNN": SVC(),
             "GradientBoosting": GradientBoostingClassifier()
         }
         results = {}
         for name, model in models.items():
             pipe = Pipeline([
                 ('scaler', StandardScaler()),
                 ('model', model)
             ])
             pipe.fit(X_train, y_train)
             y_pred = pipe.predict(X_test)
             acc = accuracy_score(y_test, y_pred)
             results[name] = acc
             print(f"(name) Accuracy: {acc:.4f}")
             print(classification_report(y_test, y_pred))
```

			, ,,	
(name) Accura	cy: 0.8177			
•	precision	recall	f1-score	support
<=50K	0.84	0.94	0.89	7028
>50K	0.71	0.45	0.55	2320
accuracy			0.82	9348
macro avg	0.77	0.70	0.72	9348
weighted avg	0.81	0.82	0.80	9348
weighted avg	0.81	0.02	0.80	3340
(name) Accura	•			
	precision	recall	f1-score	support
<=50K	0.89	0.93	0.91	7028
>50K	0.75	0.64	0.69	2320
/J0K	0.75	0.04	0.05	2320
accuracy			0.86	9348
macro avg	0.82	0.78	0.80	9348
weighted avg	0.85	0.86	0.85	9348
(name) Accura	•			
	precision	recall	f1-score	support
<=50K	0.87	0.95	0.90	7028
>50K	0.77	0.56	0.65	2320
750K	0.77	0.50	0.03	2320
accuracy			0.85	9348
macro avg	0.82	0.75	0.78	9348
weighted avg	0.84	0.85	0.84	9348
(name) Accura	•			
	precision	recall	f1-score	support
<=50K	0.88	0.95	0.91	7028
>50K	0.80	0.60	0.69	2320
>50K	0.00	0.00	0.05	2320
accuracy			0.86	9348
macro avg	0.84	0.78	0.80	9348
weighted avg	0.86	0.86	0.86	9348

```
In [44]: import matplotlib.pyplot as plt

plt.bar(results.keys(), results.values(), color='skyblue')
plt.ylabel('Accuracy Score')
plt.title('Model Comparison')
plt.xticks(rotation=45)
plt.grid(True)
plt.show()
```



```
In [45]:
         from sklearn.linear model import LogisticRegression
         from sklearn.ensemble import RandomForestClassifier, GradientBoostingClassifie
         from sklearn.neighbors import KNeighborsClassifier
         from sklearn.svm import SVC
         from sklearn.model selection import train test split
         from sklearn.metrics import accuracy_score
         import joblib
         X_train, X_test, y_train, y_test = train_test_split(x, y, test_size=0.2, rando
         models = {
             "LogisticRegression": LogisticRegression(max_iter=1000),
             "RandomForest": RandomForestClassifier(),
             "KNN": KNeighborsClassifier(),
             "SVM": SVC(),
             "GradeBoosting": GradientBoostingClassifier()
         }
         results = {}
         for name, model in models.items():
             model.fit(X_train, y_train)
             preds = model.predict(X_test)
             acc = accuracy_score(y_test, preds)
             results[name] = acc
             print(f"{name}: {acc:.4f}")
         best_model_name = max(results, key=results.get)
         best_model = models[best_model_name]
         print(f"\n Best model: {best_model_name} with accuracy {results[best_model_name]}
         joblib.dump(best_model, "best_model.pkl")
         print("Saved best model as best model.pkl")
         LogisticRegression: 0.7760
         RandomForest: 0.8542
         KNN: 0.7762
         SVM: 0.7931
         GradeBoosting: 0.8649
          Best model: GradeBoosting with accuracy 0.8649
```

Saved best model as best\_model.pkl

```
In [46]:
        %%writefile app.py
         import streamlit as st
         import pandas as pd
         import joblib
         model = joblib.load("best_model.pkl")
         st.set_page_config(page_title="Employee Salary Classification", page_icon" ",
         st.title(" Employee Salary Classification App")
         st.markdown("Predict whether an employee earns >50K or ≤50K based on input fed
         st.sidebar.header("Input Employee Details")
         age = st.sidebar.slider("Age", 18, 65, 30)
         education = st.sidebar.selectbox("Education Level", [
             "Bachelors", "Masters", "PhD", "HS-grad", "Assoc", "Some-college"
         ])
         occupation = st.sidebar.selectbox("Job Role", [
             "Tech-support", "Craft-repair", "Other-service", "Sales",
             "Exec-managerial", "Prof-speciality", "Handlers-cleaners", "Machine-op-insp
             "Adm-clerical" "Farming-fishing", "Transport-moving", "Priv-house-serv",
             "Protective-serv", "Armed-Forces"
         ])
         hours_per_week = st.sidebar.slider("Hours per week", 1, 80, 40)
         experience = st.sidebar.slider("Years of Experience", 0, 40, 5)
         input_df = pd.DataFrame({
             'age': [age],
              'education': [education],
             'occupation': [occupation],
             'hours-per-week': [hours-per-week],
             'experience':[experience]
         })
         input_df = pd.DataFrame({
             'age': [age],
             'education': [education]
             'occupation': [occupation]
             'hours-per-week': [hours-per-week]
              'experience': [experience]
         })
         st.write("### Input Class")
         st.write(input_df)
         if st.button("Predict Salary Class"):
             prediction = model.predict(input_df)
             st.success(f" Prediction: {prediction[0]}")
         st.markdown("....")
         st.markdown("#### Batch Prediction")
         uploaded_file = st.file_uploaded("Upload a CSV file for batch prediction", typ
         if uploaded file is not None:
             batch_data = pd.read_csv(uploaded_file)
             st.write("Uploaded data preview:", batch data.head())
             batch_preds = model.predict(batch_data)
             batch_data['PredictedClass'] = batch_preds
             st.write("Predictions:")
             st.write(batch data.head())
```

```
csv = batch_data.to_csv(index=False).encode('utf-8')
st.download_button("Download Preditions CSV", csv, file_name='predicted_cl
```

Overwriting app.py

```
In [47]: |y
Out[47]: 0
                   <=50K
         1
                   <=50K
         2
                    >50K
          3
                    >50K
                   <=50K
                   . . .
                   <=50K
         48837
         48838
                   >50K
         48839
                   <=50K
         48840
                   <=50K
         48841
                    >50K
         Name: income, Length: 46739, dtype: object
```

In [48]: !pip install streamlit pyngrok

e-packages (1.47.0)

Requirement already satisfied: streamlit in c:\users\lavan\anaconda3\lib\sit

```
Requirement already satisfied: pyngrok in c:\users\lavan\anaconda3\lib\site-
packages (7.2.12)
Requirement already satisfied: altair<6,>=4.0 in c:\users\lavan\anaconda3\li
b\site-packages (from streamlit) (5.5.0)
Requirement already satisfied: blinker<2,>=1.5.0 in c:\users\lavan\anaconda3
\lib\site-packages (from streamlit) (1.9.0)
Requirement already satisfied: cachetools<7,>=4.0 in c:\users\lavan\anaconda
3\lib\site-packages (from streamlit) (6.1.0)
Requirement already satisfied: click<9,>=7.0 in c:\users\lavan\anaconda3\lib
\site-packages (from streamlit) (8.0.4)
Requirement already satisfied: numpy<3,>=1.23 in c:\users\lavan\anaconda3\li
b\site-packages (from streamlit) (1.24.3)
Requirement already satisfied: packaging<26,>=20 in c:\users\lavan\anaconda3
\lib\site-packages (from streamlit) (23.1)
Requirement already satisfied: pandas<3,>=1.4.0 in c:\users\lavan\anaconda3
\lib\site-packages (from streamlit) (2.0.3)
Requirement already satisfied: pillow<12,>=7.1.0 in c:\users\lavan\anaconda3
\lib\site-packages (from streamlit) (9.4.0)
Requirement already satisfied: protobuf<7,>=3.20 in c:\users\lavan\anaconda3
\lib\site-packages (from streamlit) (6.31.1)
Requirement already satisfied: pyarrow>=7.0 in c:\users\lavan\anaconda3\lib
\site-packages (from streamlit) (11.0.0)
Requirement already satisfied: requests<3,>=2.27 in c:\users\lavan\anaconda3
\lib\site-packages (from streamlit) (2.31.0)
Requirement already satisfied: tenacity<10,>=8.1.0 in c:\users\lavan\anacond
a3\lib\site-packages (from streamlit) (8.2.2)
Requirement already satisfied: toml<2,>=0.10.1 in c:\users\lavan\anaconda3\l
ib\site-packages (from streamlit) (0.10.2)
Requirement already satisfied: typing-extensions<5,>=4.4.0 in c:\users\lavan
\anaconda3\lib\site-packages (from streamlit) (4.12.2)
Requirement already satisfied: watchdog<7,>=2.1.5 in c:\users\lavan\anaconda
3\lib\site-packages (from streamlit) (2.1.6)
Requirement already satisfied: gitpython!=3.1.19,<4,>=3.0.7 in c:\users\lava
n\anaconda3\lib\site-packages (from streamlit) (3.1.44)
Requirement already satisfied: pydeck<1,>=0.8.0b4 in c:\users\lavan\anaconda
3\lib\site-packages (from streamlit) (0.9.1)
Requirement already satisfied: tornado!=6.5.0,<7,>=6.0.3 in c:\users\lavan\a
naconda3\lib\site-packages (from streamlit) (6.3.2)
Requirement already satisfied: PyYAML>=5.1 in c:\users\lavan\anaconda3\lib\s
ite-packages (from pyngrok) (6.0)
Requirement already satisfied: jinja2 in c:\users\lavan\anaconda3\lib\site-p
ackages (from altair<6,>=4.0->streamlit) (3.1.2)
Requirement already satisfied: jsonschema>=3.0 in c:\users\lavan\anaconda3\l
ib\site-packages (from altair<6,>=4.0->streamlit) (4.17.3)
Requirement already satisfied: narwhals>=1.14.2 in c:\users\lavan\anaconda3
\lib\site-packages (from altair<6,>=4.0->streamlit) (1.47.1)
Requirement already satisfied: colorama in c:\users\lavan\anaconda3\lib\site
-packages (from click<9,>=7.0->streamlit) (0.4.6)
Requirement already satisfied: gitdb<5,>=4.0.1 in c:\users\lavan\anaconda3\l
ib\site-packages (from gitpython!=3.1.19,<4,>=3.0.7->streamlit) (4.0.12)
Requirement already satisfied: python-dateutil>=2.8.2 in c:\users\lavan\anac
onda3\lib\site-packages (from pandas<3,>=1.4.0->streamlit) (2.8.2)
Requirement already satisfied: pytz>=2020.1 in c:\users\lavan\anaconda3\lib
\site-packages (from pandas<3,>=1.4.0->streamlit) (2023.3.post1)
Requirement already satisfied: tzdata>=2022.1 in c:\users\lavan\anaconda3\li
b\site-packages (from pandas<3,>=1.4.0->streamlit) (2023.3)
Requirement already satisfied: charset-normalizer<4,>=2 in c:\users\lavan\an
aconda3\lib\site-packages (from requests<3,>=2.27->streamlit) (2.0.4)
Requirement already satisfied: idna<4,>=2.5 in c:\users\lavan\anaconda3\lib
```

```
\site-packages (from requests<3,>=2.27->streamlit) (3.4)
Requirement already satisfied: urllib3<3,>=1.21.1 in c:\users\lavan\anaconda
3\lib\site-packages (from requests<3,>=2.27->streamlit) (1.26.16)
Requirement already satisfied: certifi>=2017.4.17 in c:\users\lavan\anaconda
3\lib\site-packages (from requests<3,>=2.27->streamlit) (2023.7.22)
Requirement already satisfied: smmap<6,>=3.0.1 in c:\users\lavan\anaconda3\l
ib\site-packages (from gitdb<5,>=4.0.1->gitpython!=3.1.19,<4,>=3.0.7->stream
lit) (5.0.2)
Requirement already satisfied: MarkupSafe>=2.0 in c:\users\lavan\anaconda3\l
ib\site-packages (from jinja2->altair<6,>=4.0->streamlit) (2.1.1)
Requirement already satisfied: attrs>=17.4.0 in c:\users\lavan\anaconda3\lib
\site-packages (from jsonschema>=3.0->altair<6,>=4.0->streamlit) (24.2.0)
Requirement already satisfied: pyrsistent!=0.17.0,!=0.17.1,!=0.17.2,>=0.14.0
in c:\users\lavan\anaconda3\lib\site-packages (from jsonschema>=3.0->altair
6,>=4.0->streamlit) (0.18.0)
Requirement already satisfied: six>=1.5 in c:\users\lavan\anaconda3\lib\site
-packages (from python-dateutil>=2.8.2->pandas<3,>=1.4.0->streamlit) (1.16.
0)
```

```
In [49]: import os
import threading

def run_streamlit():
    os.system('streamlit run app.py --server.port 8501')

    thread = threading.Thread(target=run_streamlit)
    thread.start()
```

```
In [50]: from pyngrok import ngrok
import time

time.sleep(5)

public_url = ngrok.connect(8501)
print("Your Streamlit app is live here:",public_url)
```

Installing ngrok ...

```
OSError
                                          Traceback (most recent call last)
File ~\anaconda3\Lib\site-packages\pyngrok\installer.py:192, in install_ngro
k(ngrok_path, ngrok_version, **kwargs)
            download_path = _download_file(url, **kwargs)
    190
--> 192
            _install_ngrok_zip(ngrok_path, download_path)
    193 except Exception as e:
File ~\anaconda3\Lib\site-packages\pyngrok\installer.py:207, in _install_ngr
ok_zip(ngrok_path, zip_path)
    205 _print_progress("Installing ngrok ... ")
--> 207 with zipfile.ZipFile(zip_path, "r") as zip_ref:
          logger.debug(f"Extracting ngrok binary from {zip path} to {ngrok
_path} ...")
File ~\anaconda3\Lib\zipfile.py:1284, in ZipFile.__init__(self, file, mode,
compression, allowZip64, compresslevel, strict_timestamps, metadata_encodin
g)
   1283 trv:
-> 1284
          self.fp = io.open(file, filemode)
   1285 except OSError:
OSError: [Errno 22] Invalid argument: 'C:\\Users\\lavan\\AppData\\Local\\Tem
p\\ngrok-v3-stable-windows-amd64.zip'
During handling of the above exception, another exception occurred:
PyngrokNgrokInstallError
                                          Traceback (most recent call last)
Cell In[50], line 6
      2 import time
      4 time.sleep(5)
---> 6 public url = ngrok.connect(8501)
      7 print("Your Streamlit app is live here:",public_url)
File ~\anaconda3\Lib\site-packages\pyngrok\ngrok.py:385, in connect(addr, pr
oto, name, pyngrok_config, **options)
    381 upgrade legacy params(pyngrok config, options)
    383 logger.info(f"Opening tunnel named: {name}")
--> 385 api_url = get_ngrok_process(pyngrok_config).api_url
    387 logger.debug(f"Creating tunnel with options: {options}")
    389 tunnel = NgrokTunnel(api_request(f"{api_url}/api/tunnels", method="P
OST", data=options,
    390
                                         timeout=pyngrok config.request time
out),
                             pyngrok_config, api_url)
    391
File ~\anaconda3\Lib\site-packages\pyngrok\ngrok.py:201, in get_ngrok_proces
s(pyngrok_config)
    198 if pyngrok config is None:
            pyngrok config = conf.get default()
--> 201 install_ngrok(pyngrok_config)
    203 return process.get_process(pyngrok_config)
File ~\anaconda3\Lib\site-packages\pyngrok\ngrok.py:129, in install_ngrok(py
ngrok config)
    126
            pyngrok_config = conf.get_default()
    128 if not os.path.exists(pyngrok config.ngrok path):
            installer.install_ngrok(pyngrok_config.ngrok_path, ngrok_version
--> 129
=pyngrok_config.ngrok_version)
    131 config_path = conf.get_config_path(pyngrok_config)
    133 # Install the config to the requested path
```

```
File ~\anaconda3\Lib\site-packages\pyngrok\installer.py:194, in install_ngro
        k(ngrok_path, ngrok_version, **kwargs)
                    _install_ngrok_zip(ngrok_path, download_path)
            192
            193 except Exception as e:
                    raise PyngrokNgrokInstallError(f"An error occurred while downloa
        --> 194
        ding ngrok from {url}: {e}")
        PyngrokNgrokInstallError: An error occurred while downloading ngrok from htt
        ps://bin.equinox.io/c/bNyj1mQVY4c/ngrok-v3-stable-windows-amd64.zip: (http
        s://bin.equinox.io/c/bNyj1mQVY4c/ngrok-v3-stable-windows-amd64.zip:) [Errno
        22] Invalid argument: 'C:\\Users\\lavan\\AppData\\Local\\Temp\\ngrok-v3-stab
        le-windows-amd64.zip'
In [ ]: | from sklearn.preprocessing import MinMaxScaler
        scaler=MinMaxScaler()
        x=scaler.fit_transform(x)
        Χ
In [ ]: from sklearn.model_selection import train_test_split
        xtrain, xtest, ytrain, ytest= train_test_split(x,y, test_size=0.2, random_stat
In [ ]: xtrain
In [ ]: #machine learning algorithm
        from sklearn.neighbors import KNeighborsClassifier
        knn=KNeighborsClassifier()
        knn.fit(xtrain, ytrain)
        predict=knn.predict(xtest)
        predict
In [ ]: from sklearn.metrics import accuracy_score
        accuracy score(ytest,predict)
In [ ]: | from sklearn.linear model import LogisticRegression
        lr=LogisticRegression()
        lr.fit(xtrain, ytrain)
        predict1=lr.predict(xtest)
        predict1
In [ ]: from sklearn.metrics import accuracy score
        accuracy score(ytest,predict)
In [ ]: from sklearn.neural network import MLPClassifier
        clf=MLPClassifier(solver='adam', hidden_layer_sizes=(5,2), random_state=2, max
        clf.fit(xtrain, ytrain)
        predict2=clf.predict(xtest)
        predict2
```

In [ ]:	<pre>from sklearn.metrics import accuracy_score accuracy_score(ytest,predict2)</pre>
In [ ]:	
In [ ]:	
In [ ]:	