

✓ 1) IMPORT LIBRARIES


```
!pip install gradio
```

 [Show hidden output](#)


```
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LogisticRegression
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import accuracy_score, confusion_matrix, classification_report
```

✓ 2) LOAD THE DATASET

```
from google.colab import files
uploaded = files.upload()
```

 No file chosen Upload widget is only available when the cell has been executed in the current browser session. Please rerun this cell to enable.

```
df = pd.read_csv("diabetes.csv")
print(df.head())
```



	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	BMI	\
0	6	148	72	35	0	33.6	
1	1	85	66	29	0	26.6	
2	8	183	64	0	0	23.3	
3	1	89	66	23	94	28.1	
4	0	137	40	35	168	43.1	

	DiabetesPedigreeFunction	Age	Outcome
0	0.627	50	1
1	0.351	31	0
2	0.672	32	1
3	0.167	21	0
4	2.288	33	1

✓ 3) EXPLORE THE DATA

```
print(df.info())
```

```
print(df.describe())
```

```
# Correlation heatmap
plt.figure(figsize=(8,6))
sns.heatmap(df.corr(), annot=True, cmap="coolwarm")
plt.title("Feature Correlation Heatmap")
plt.show()
```

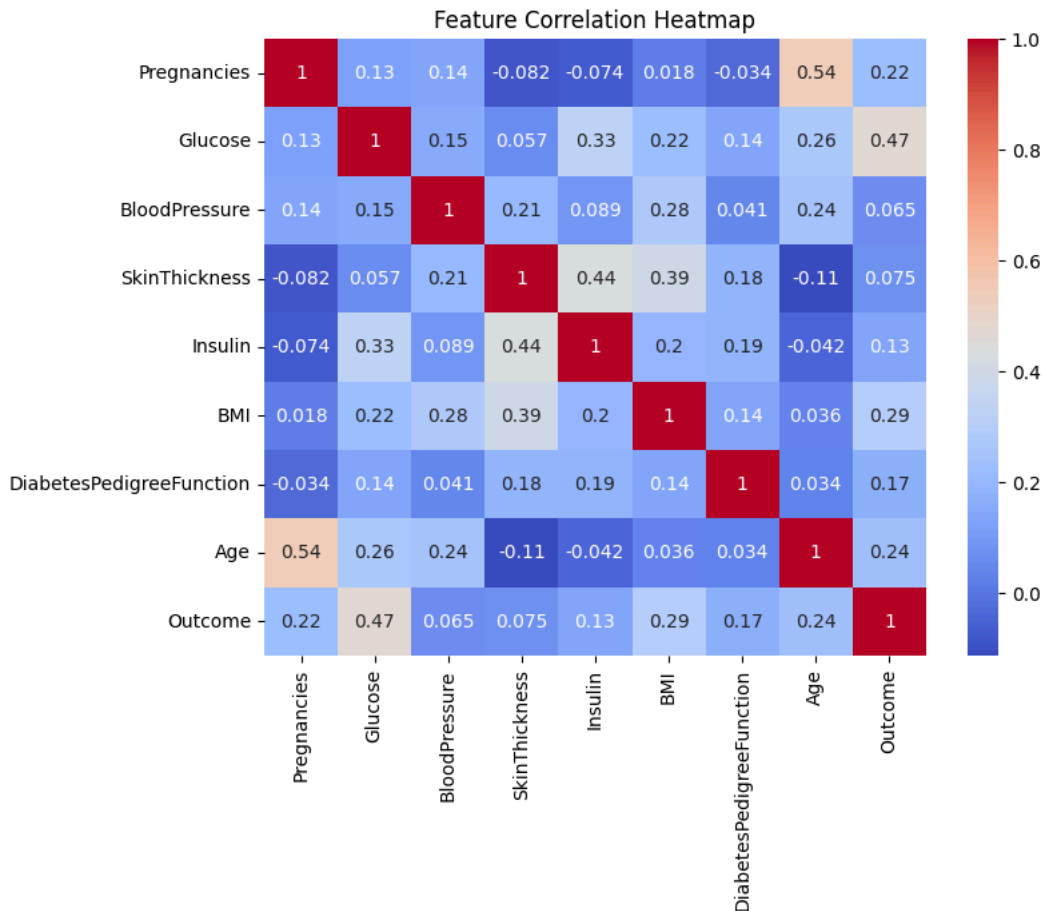
```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 768 entries, 0 to 767
Data columns (total 9 columns):
#   Column                Non-Null Count  Dtype
---  -
0   Pregnancies            768 non-null   int64
1   Glucose                768 non-null   int64
2   BloodPressure          768 non-null   int64
3   SkinThickness          768 non-null   int64
4   Insulin                768 non-null   int64
5   BMI                   768 non-null   float64
6   DiabetesPedigreeFunction 768 non-null   float64
7   Age                   768 non-null   int64
8   Outcome               768 non-null   int64
dtypes: float64(2), int64(7)
memory usage: 54.1 KB
None

```

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin
count	768.000000	768.000000	768.000000	768.000000	768.000000
mean	3.845052	120.894531	69.105469	20.536458	79.799479
std	3.369578	31.972618	19.355807	15.952218	115.244002
min	0.000000	0.000000	0.000000	0.000000	0.000000
25%	1.000000	99.000000	62.000000	0.000000	0.000000
50%	3.000000	117.000000	72.000000	23.000000	30.500000
75%	6.000000	140.250000	80.000000	32.000000	127.250000
max	17.000000	199.000000	122.000000	99.000000	846.000000

	BMI	DiabetesPedigreeFunction	Age	Outcome
count	768.000000	768.000000	768.000000	768.000000
mean	31.992578	0.471876	33.240885	0.348958
std	7.884160	0.331329	11.760232	0.476951
min	0.000000	0.078000	21.000000	0.000000
25%	27.300000	0.243750	24.000000	0.000000
50%	32.000000	0.372500	29.000000	0.000000
75%	36.600000	0.626250	41.000000	1.000000
max	67.100000	2.420000	81.000000	1.000000



5) SPLIT THE DATA (TRAIN/TEST)

```

df = pd.read_csv("diabetes.csv")

X = df[['Age', 'Glucose', 'BloodPressure']]
y = df['BMI']

X_train, X_test, y_train, y_test = train_test_split(
    X, y, test_size=0.2, random_state=42
)

print("Training features:\n", X_train)
print("Test features:\n", X_test)
print("Training labels:\n", y_train)
print("Test labels:\n", y_test)

```

↩ Training features:

	Age	Glucose	BloodPressure
60	21	84	0
618	50	112	82
346	22	139	46
294	65	161	50
231	46	134	80
..
71	26	139	64
106	27	96	122
270	38	101	86
435	29	141	0
102	21	125	96

[614 rows x 3 columns]

Test features:

	Age	Glucose	BloodPressure
668	43	98	58
324	21	112	75
624	21	108	64
690	34	107	80
473	50	136	90
..
355	49	165	88
534	24	77	56
344	57	95	72
296	29	146	70
462	39	74	70

[154 rows x 3 columns]

Training labels:

60	0.0
618	28.2
346	28.7
294	21.9
231	46.2
..	...
71	28.6
106	22.4
270	45.6
435	42.4
102	22.5

Name: BMI, Length: 614, dtype: float64

Test labels:

668	34.0
324	35.7
624	30.8
690	24.6
473	29.9
..	...
355	30.4
534	33.3
344	36.8
296	28.0
462	35.3

Name: BMI, Length: 154, dtype: float64

6) REGRESSION MODEL

6) [A] LOGISTIC REGRESION MODEL

```
# Train logistic regression model
log_model = LogisticRegression(max_iter=1000)
log_model.fit(X_train, y_train)

# Predict
log_preds = log_model.predict(X_test)

# Evaluate
print("Logistic Regression Accuracy:", accuracy_score(y_test, log_preds))
print("Classification Report:\n", classification_report(y_test, log_preds))
```

```
Logistic Regression Accuracy: 0.7467532467532467
Classification Report:
              precision    recall  f1-score   support

     0       0.81         0.79         0.80         99
     1       0.64         0.67         0.65         55

 accuracy          0.75         0.75         0.75         154
 macro avg         0.73         0.73         0.73         154
 weighted avg      0.75         0.75         0.75         154
```

6) [B] RANDOM FOREST MODEL

```
rf_model = RandomForestClassifier(n_estimators=100, random_state=42)
rf_model.fit(X_train, y_train)

rf_preds = rf_model.predict(X_test)

print("Random Forest Accuracy:", accuracy_score(y_test, rf_preds))
print("Classification Report:\n", classification_report(y_test, rf_preds))
```

```
Random Forest Accuracy: 0.7207792207792207
Classification Report:
              precision    recall  f1-score   support

     0       0.79         0.78         0.78         99
     1       0.61         0.62         0.61         55

 accuracy          0.72         0.72         0.72         154
 macro avg         0.70         0.70         0.70         154
 weighted avg      0.72         0.72         0.72         154
```

7) MAKE PREDICTIONS

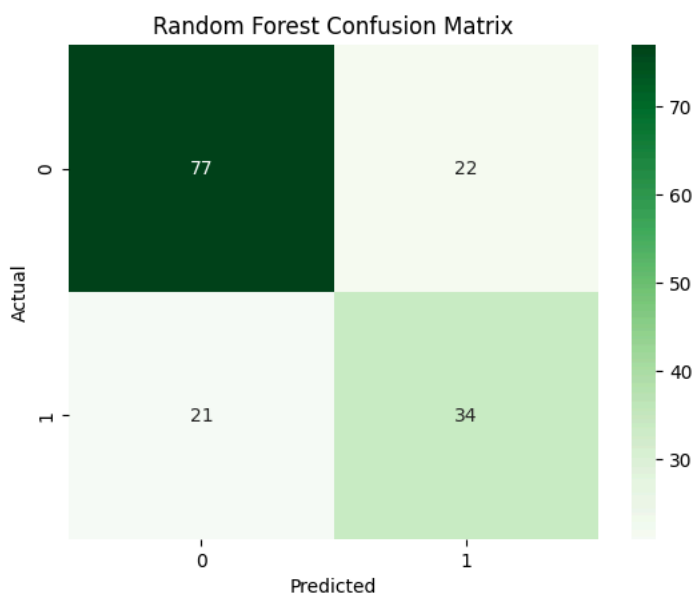
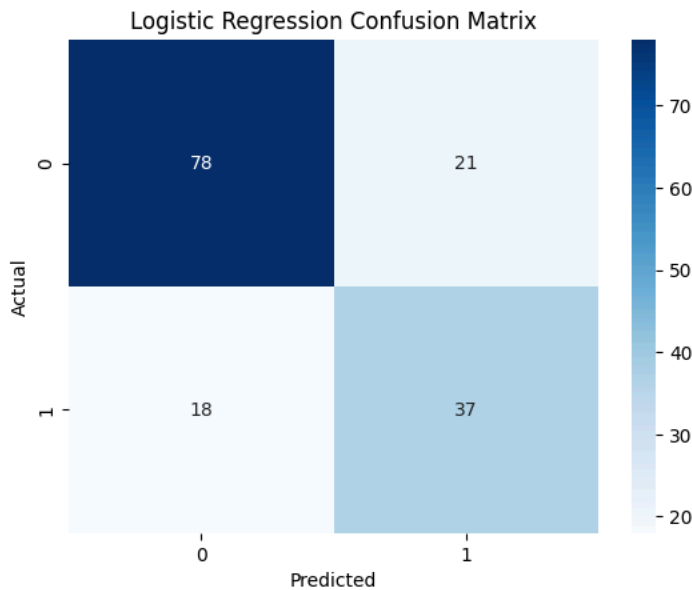
```
# Predict on test data
y_pred = model.predict(X_test)
```

8) EVALUATE THE MODEL

```
# Logistic Regression Confusion Matrix
sns.heatmap(confusion_matrix(y_test, log_preds), annot=True, fmt='d', cmap='Blues')
plt.title("Logistic Regression Confusion Matrix")
plt.xlabel("Predicted")
plt.ylabel("Actual")
plt.show()
```

```
# Random Forest Confusion Matrix
```

```
sns.heatmap(confusion_matrix(y_test, rf_preds), annot=True, fmt='d', cmap='Greens')
plt.title("Random Forest Confusion Matrix")
plt.xlabel("Predicted")
plt.ylabel("Actual")
plt.show()
```



▼ Build The Model

```
import gradio as gr
import numpy as np

from sklearn.linear_model import LogisticRegression
from sklearn.datasets import make_classification
from sklearn.model_selection import train_test_split

# Sample data and model (for demo purposes)
X, y = make_classification(n_samples=500, n_features=5, random_state=42)
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
model = LogisticRegression()
model.fit(X_train, y_train)

# Prediction function
```

```
def predict_health_risk(age, gender, bmi, blood_pressure, glucose):
    gender_val = 1 if gender == "Male" else 0
    features = np.array([[age, gender_val, bmi, blood_pressure, glucose]])
    pred = model.predict(features)[0]
    return "High Risk" if pred == 1 else "Low Risk"

# Gradio Interface
iface = gr.Interface(
    fn=predict_health_risk,
    inputs=[
        gr.Number(label="Age"),
        gr.Radio(["Male", "Female"], label="Gender"),
        gr.Number(label="BMI"),
        gr.Number(label="BloodPressure"),
        gr.Number(label="Glucose"),
    ],
    outputs=gr.Text(label="Risk Prediction"),
    title="AI-Powered Health Risk Predictor",
    description="Enter patient details to predict health risk using AI.",
    theme="default"
)

# Launch app
iface.launch()
```

🔗 It looks like you are running Gradio on a hosted a Jupyter notebook. For the Gradio app to work, sharing must be enabled. Automatically

Colab notebook detected. To show errors in colab notebook, set debug=True in launch()

* Running on public URL: <https://36587948784983e87a.gradio.live>

This share link expires in 1 week. For free permanent hosting and GPU upgrades, run `gradio deploy` from the terminal in the working dir

AI-Powered Health Risk Predictor

Enter patient details to predict health risk using AI.

<div>Age</div> <div><input type="text" value="0"/></div>	<div>Risk Prediction</div> <div><input type="text"/></div>
<div>Gender</div> <div><input type="radio"/> Male <input type="radio"/> Female</div>	<div>Flag</div>
<div>BMI</div> <div><input type="text" value="0"/></div>	
<div>BloodPressure</div> <div><input type="text" value="0"/></div>	
<div>Glucose</div> <div><input type="text"/></div>	