```
import pandas as pd
from google.colab import files
uploaded= files.upload()
df=pd.read_csv("accident ag.csv")
print(df)
print(df.isnull().sum())
```

Choose files | accident ag.csv

• accident ag.csv(text/csv) - 4542 bytes, last modified: 15/05/2025 - 100% done Saving accident ag.csv to accident ag.csv

| | Age | Gender | <pre>Speed_of_Impact</pre> | Helmet_Used | Seatbelt_Used | Survived |
|-----|-----|--------|----------------------------|-------------|---------------|----------|
| 0 | 56 | Female | 27.0 | No | No | 1 |
| 1 | 69 | Female | 46.0 | No | Yes | 1 |
| 2 | 46 | Male | 46.0 | Yes | Yes | 0 |
| 3 | 32 | Male | 117.0 | No | Yes | 0 |
| 4 | 60 | Female | 40.0 | Yes | Yes | 0 |
| | | | • • • | | • • • | |
| 195 | 69 | Female | 111.0 | No | Yes | 1 |
| 196 | 30 | Female | 51.0 | No | Yes | 1 |
| 197 | 58 | Male | 110.0 | No | Yes | 1 |
| 198 | 20 | Male | 103.0 | No | Yes | 1 |
| 199 | 56 | Female | 43.0 | No | Yes | 1 |

```
[200 rows x 6 columns]
Age
Gender
                   1
Speed_of_Impact
                   3
Helmet_Used
                   0
Seatbelt_Used
                   0
Survived
dtype: int64
```

label encoders = {}

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.model selection import train test split
from sklearn.ensemble import RandomForestClassifier
from sklearn.preprocessing import LabelEncoder
from sklearn.metrics import classification report, accuracy score, confusion matrix
# Load dataset
df = pd.read_csv('accident ag.csv')
# Display basic info
print("Dataset Info:")
print(df.info())
print("\nMissing values:\n", df.isnull().sum())
# Fill or drop missing values (basic approach)
df = df.dropna() # or you can fillna() based on strategy
# Encode categorical features
```

```
for column in df.select_dtypes(include=['object']).columns:
    le = LabelEncoder()
    df[column] = le.fit_transform(df[column])
    label_encoders[column] = le
# Check correlation to find a target (assume last column is target if unknown)
print("\nCorrelation with target:")
print(df.corr())
# For demo, let's assume the last column is the target
X = df.iloc[:, :-1]
y = df.iloc[:, -1]
# Train/test split
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
# Random Forest Classifier
model = RandomForestClassifier(n_estimators=100, random_state=42)
model.fit(X_train, y_train)
# Prediction
y_pred = model.predict(X_test)
# Evaluation
print("\nAccuracy:", accuracy_score(y_test, y_pred))
print("\nClassification Report:\n", classification_report(y_test, y_pred))
print("\nConfusion Matrix:\n", confusion_matrix(y_test, y_pred))
# Feature importance
importances = model.feature_importances_
indices = np.argsort(importances)[::-1]
# Plot feature importances
plt.figure(figsize=(12, 6))
sns.barplot(x=importances[indices], y=X.columns[indices])
plt.title("Feature Importances")
plt.show()
```

```
→ Dataset Info:
```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 200 entries, 0 to 199
Data columns (total 6 columns):

Column Non-Null Count Dtype --- ----------0 Age 200 non-null int64 1 Gender 199 non-null object Speed_of_Impact 197 non-null float64 2 3 Helmet_Used 200 non-null object 4 Seatbelt Used 200 non-null object 5 Survived 200 non-null int64

dtypes: float64(1), int64(2), object(3)

memory usage: 9.5+ KB

None

Missing values:

Age 0
Gender 1
Speed_of_Impact 3
Helmet_Used 0
Seatbelt_Used 0
Survived 0
dtype: int64

Correlation with target:

| | Age | Gender | Speed_of_Impact | Helmet_Used | ' |
|-----------------|-----------|-----------|-----------------|-------------|---|
| Age | 1.000000 | -0.049222 | 0.106833 | 0.104798 | |
| Gender | -0.049222 | 1.000000 | -0.047972 | 0.012760 | |
| Speed_of_Impact | 0.106833 | -0.047972 | 1.000000 | -0.010659 | |
| Helmet_Used | 0.104798 | 0.012760 | -0.010659 | 1.000000 | |
| Seatbelt_Used | -0.016214 | -0.058491 | -0.000240 | 0.099879 | |
| Survived | 0.119213 | 0.115288 | 0.042902 | -0.053661 | |

Seatbelt_UsedSurvivedAge-0.0162140.119213Gender-0.0584910.115288Speed_of_Impact-0.0002400.042902Helmet_Used0.099879-0.053661Seatbelt_Used1.0000000.059183Survived0.0591831.000000

Accuracy: 0.425

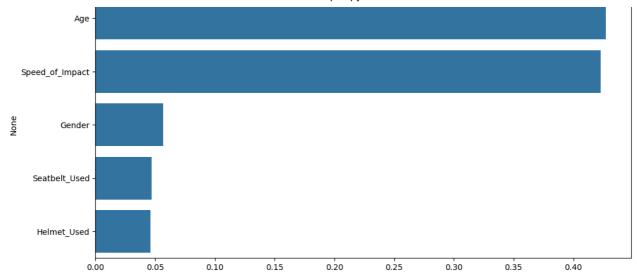
Classification Report:

| | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|---------|
| 0 | 0.58 | 0.28 | 0.38 | 25 |
| 1 | 0.36 | 0.67 | 0.47 | 15 |
| accuracy | | | 0.42 | 40 |
| macro avg | 0.47 | 0.47 | 0.42 | 40 |
| weighted avg | 0.50 | 0.42 | 0.41 | 40 |

Confusion Matrix:

[[7 18] [5 10]]

Feature Importances



```
!pip install gradio
import pandas as pd
import numpy as np
import gradio as gr
from sklearn.ensemble import RandomForestClassifier
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import LabelEncoder
import joblib
# Load data
df = pd.read_csv("accident (1).csv")
# Preprocess
df = df.dropna()
# Label Encoding
label_encoders = {}
for col in df.select_dtypes(include='object').columns:
    le = LabelEncoder()
    df[col] = le.fit_transform(df[col])
    label_encoders[col] = le
# Features and target (you can customize target column here)
X = df.iloc[:, :-1]
y = df.iloc[:, -1]
# Train/test split
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_sta
# Train model
model = RandomForestClassifier(n_estimators=100, random_state=42)
model.fit(X_train, y_train)
# Save model
joblib.dump(model, "accident_model.pkl")
# Gradio interface
def predict_accident(*inputs):
```

if __name__ == "__main__":
 demo.launch()

```
→ Collecting gradio
      Downloading gradio-5.29.1-py3-none-any.whl.metadata (16 kB)
    Collecting aiofiles<25.0,>=22.0 (from gradio)
      Downloading aiofiles-24.1.0-py3-none-any.whl.metadata (10 kB)
    Requirement already satisfied: anyio<5.0,>=3.0 in /usr/local/lib/python3.11/dist-pack
    Collecting fastapi<1.0,>=0.115.2 (from gradio)
      Downloading fastapi-0.115.12-py3-none-any.whl.metadata (27 kB)
    Collecting ffmpy (from gradio)
      Downloading ffmpy-0.5.0-py3-none-any.whl.metadata (3.0 kB)
    Collecting gradio-client==1.10.1 (from gradio)
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    Collecting groovy~=0.1 (from gradio)
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    Requirement already satisfied: markupsafe<4.0,>=2.0 in /usr/local/lib/python3.11/dist
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    Requirement already satisfied: pandas<3.0,>=1.0 in /usr/local/lib/python3.11/dist-pac
    Requirement already satisfied: pillow<12.0,>=8.0 in /usr/local/lib/python3.11/dist-pa
    Requirement already satisfied: pydantic<2.12,>=2.0 in /usr/local/lib/python3.11/dist-
    Collecting pydub (from gradio)
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    Collecting python-multipart>=0.0.18 (from gradio)
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    Requirement already satisfied: pyyaml<7.0,>=5.0 in /usr/local/lib/python3.11/dist-pac
    Collecting ruff>=0.9.3 (from gradio)
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    Collecting safehttpx<0.2.0,>=0.1.6 (from gradio)
      Downloading safehttpx-0.1.6-py3-none-any.whl.metadata (4.2 kB)
    Collecting semantic-version~=2.0 (from gradio)
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    Collecting starlette<1.0,>=0.40.0 (from gradio)
      Downloading starlette-0.46.2-py3-none-any.whl.metadata (6.2 kB)
    Collecting tomlkit<0.14.0,>=0.12.0 (from gradio)
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    Requirement already satisfied: typing-extensions~=4.0 in /usr/local/lib/python3.11/di
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    Requirement already satisfied: certifi in /usr/local/lib/python3.11/dist-packages (fr
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    Requirement already satisfied: filelock in /usr/local/lib/python3.11/dist-packages (f
    Requirement already satisfied: requests in /usr/local/lib/python3.11/dist-packages (f
    Requirement already satisfied: tqdm>=4.42.1 in /usr/local/lib/python3.11/dist-package
    Requirement already satisfied: hf-xet<2.0.0,>=1.1.0 in /usr/local/lib/python3.11/dist
    Requirement already satisfied: python-dateutil>=2.8.2 in /usr/local/lib/python3.11/di
    Requirement already satisfied: pytz>=2020.1 in /usr/local/lib/python3.11/dist-package
    Requirement already satisfied: tzdata>=2022.7 in /usr/local/lib/python3.11/dist-packa
```

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