```
import pandas as pd
df = pd.read csv('creditcard.csv.zip')
df.head()
{"type": "dataframe", "variable name": "df"}
print("[ Dataset Info:\n")
print(df.info())
print("\n□ Missing Values:\n")
print(df.isnull().sum())
print("\n[ Class Distribution:\n")
print(df['Class'].value counts())
□ Dataset Info:
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 284807 entries, 0 to 284806
Data columns (total 31 columns):
     Column Non-Null Count
                              Dtype
                              - - - - -
 0
     Time
             284807 non-null float64
 1
     ٧1
             284807 non-null float64
 2
     ٧2
             284807 non-null float64
 3
    ٧3
             284807 non-null float64
 4
     ۷4
             284807 non-null float64
 5
     ۷5
             284807 non-null float64
 6
     ۷6
             284807 non-null float64
             284807 non-null float64
 7
    ٧7
 8
    8V
             284807 non-null float64
 9
     ٧9
             284807 non-null float64
 10
    V10
             284807 non-null float64
 11
    V11
             284807 non-null float64
12
    V12
             284807 non-null float64
 13
    V13
             284807 non-null float64
 14
    V14
             284807 non-null float64
 15
    V15
             284807 non-null float64
 16
    V16
             284807 non-null float64
 17
    V17
             284807 non-null float64
             284807 non-null float64
 18
    V18
 19
    V19
             284807 non-null float64
 20
    V20
             284807 non-null float64
 21
    V21
             284807 non-null float64
22
    V22
             284807 non-null float64
             284807 non-null float64
 23
    V23
             284807 non-null float64
 24
    V24
 25
    V25
             284807 non-null float64
```

```
26 V26
27 V27
             284807 non-null float64
             284807 non-null float64
 28 V28
             284807 non-null float64
29 Amount 284807 non-null float64
30 Class 284807 non-null int64
dtypes: float64(30), int64(1)
memory usage: 67.4 MB
None

☐ Missing Values:
Time
          0
          0
٧1
٧2
          0
٧3
          0
٧4
          0
۷5
          0
          0
V6
V7
          0
8
          0
۷9
          0
V10
          0
V11
          0
V12
          0
          0
V13
V14
          0
V15
          0
V16
          0
V17
          0
          0
V18
V19
          0
V20
          0
V21
          0
V22
          0
V23
          0
V24
          0
V25
          0
V26
          0
V27
          0
V28
          0
Amount
          0
Class
dtype: int64
☐ Class Distribution:
Class
0
     284315
1
        492
Name: count, dtype: int64
```

```
from sklearn.model selection import train_test_split
from sklearn.preprocessing import StandardScaler
from imblearn.over sampling import SMOTE
X = df.drop('Class', axis=1)
y = df['Class']
scaler = StandardScaler()
X scaled = scaler.fit transform(X)
smote = SMOTE(random state=42)
X resampled, y resampled = smote.fit resample(X scaled, y)
print("□ Class distribution before SMOTE:")
print(y.value counts())
print("\n□ Class distribution after SMOTE:")
print(pd.Series(y resampled).value counts())
☐ Class distribution before SMOTE:
Class
0
    284315
1
        492
Name: count, dtype: int64

□ Class distribution after SMOTE:

Class
    284315
0
1
     284315
Name: count, dtype: int64
from sklearn.linear model import LogisticRegression
from sklearn.tree import DecisionTreeClassifier
from sklearn.neural network import MLPClassifier
from sklearn.metrics import confusion matrix, classification report
from sklearn.model_selection import train_test_split
X train, X test, y train, y test = train test split(X resampled,
y_resampled,
                                                     test size=0.2,
random state=42)
lr = LogisticRegression(max iter=1000)
lr.fit(X train, y_train)
lr_pred = lr.predict(X_test)
print("□ Logistic Regression Report:\n")
```

```
print(confusion matrix(y test, lr pred))
print(classification report(y test, lr pred))
☐ Logistic Regression Report:
[[55361 1389]
 [ 4289 52687]]
              precision
                            recall f1-score
                                                support
                    0.93
                              0.98
                                         0.95
           0
                                                  56750
           1
                    0.97
                              0.92
                                         0.95
                                                  56976
                                         0.95
                                                 113726
    accuracy
   macro avq
                    0.95
                              0.95
                                         0.95
                                                 113726
weighted avg
                    0.95
                              0.95
                                         0.95
                                                 113726
dt = DecisionTreeClassifier(random state=42)
dt.fit(X train, y train)
dt_pred = dt.predict(X_test)
print("[ Decision Tree Report:\n")
print(confusion matrix(y test, dt pred))
print(classification report(y test, dt pred))
□ Decision Tree Report:
[[56605]]
          1451
     58 56918]]
                            recall f1-score
              precision
                                                support
           0
                    1.00
                              1.00
                                         1.00
                                                  56750
           1
                    1.00
                              1.00
                                         1.00
                                                  56976
                                         1.00
                                                 113726
    accuracy
                    1.00
                              1.00
                                         1.00
   macro avg
                                                 113726
weighted avg
                    1.00
                              1.00
                                         1.00
                                                 113726
mlp = MLPClassifier(hidden layer sizes=(100,), max iter=300,
random state=42)
mlp.fit(X_train, y_train)
mlp pred = mlp.predict(X test)
print("□ Neural Network Report:\n")
print(confusion_matrix(y_test, mlp_pred))
print(classification report(y test, mlp pred))

    □ Neural Network Report:

[[56721
           291
```

```
0 5697611
                            recall f1-score
              precision
                                                 support
                    1.00
                               1.00
                                         1.00
                                                   56750
           1
                    1.00
                               1.00
                                         1.00
                                                   56976
                                         1.00
    accuracy
                                                  113726
                                         1.00
                    1.00
                               1.00
                                                  113726
   macro avq
weighted avg
                    1.00
                              1.00
                                         1.00
                                                  113726
from sklearn.ensemble import IsolationForest
from sklearn.metrics import classification report, confusion matrix
iso_forest = IsolationForest(contamination=0.01, random state=42)
y_pred_iforest = iso_forest.fit_predict(X_scaled)
y pred iforest = [1 \text{ if } x == -1 \text{ else } 0 \text{ for } x \text{ in } y \text{ pred iforest}]
print("□ Isolation Forest Results:")
print(confusion matrix(y, y pred iforest))
print(classification report(y, y pred iforest))

    □ Isolation Forest Results:

[[281755
           2560]
     203
            28911
               precision
                            recall f1-score
                                                 support
                              0.99
           0
                    1.00
                                         1.00
                                                  284315
           1
                    0.10
                              0.59
                                         0.17
                                                     492
                                         0.99
                                                  284807
    accuracy
                    0.55
                              0.79
                                         0.58
   macro avq
                                                  284807
                              0.99
weighted avg
                    1.00
                                         0.99
                                                 284807
!pip install tensorflow --upgrade
Requirement already satisfied: tensorflow in
/usr/local/lib/python3.11/dist-packages (2.18.0)
Collecting tensorflow
  Downloading tensorflow-2.19.0-cp311-cp311-
manylinux_2_17_x86_64.manylinux2014_x86_64.whl.metadata (4.1 kB)
Requirement already satisfied: absl-py>=1.0.0 in
/usr/local/lib/python3.11/dist-packages (from tensorflow) (1.4.0)
Requirement already satisfied: astunparse>=1.6.0 in
/usr/local/lib/python3.11/dist-packages (from tensorflow) (1.6.3)
Requirement already satisfied: flatbuffers>=24.3.25 in
/usr/local/lib/python3.11/dist-packages (from tensorflow) (25.2.10)
```

```
Reguirement already satisfied: gast!=0.5.0,!=0.5.1,!=0.5.2,>=0.2.1
in /usr/local/lib/python3.11/dist-packages (from tensorflow) (0.6.0)
Requirement already satisfied: google-pasta>=0.1.1 in
/usr/local/lib/python3.11/dist-packages (from tensorflow) (0.2.0)
Requirement already satisfied: libclang>=13.0.0 in
/usr/local/lib/python3.11/dist-packages (from tensorflow) (18.1.1)
Requirement already satisfied: opt-einsum>=2.3.2 in
/usr/local/lib/python3.11/dist-packages (from tensorflow) (3.4.0)
Requirement already satisfied: packaging in
/usr/local/lib/python3.11/dist-packages (from tensorflow) (24.2)
Requirement already satisfied: protobuf!=4.21.0,!=4.21.1,!=4.21.2,!
=4.21.3,!=4.21.4,!=4.21.5,<6.0.0dev,>=3.20.3 in
/usr/local/lib/python3.11/dist-packages (from tensorflow) (5.29.5)
Requirement already satisfied: requests<3,>=2.21.0 in
/usr/local/lib/python3.11/dist-packages (from tensorflow) (2.32.3)
Requirement already satisfied: setuptools in
/usr/local/lib/python3.11/dist-packages (from tensorflow) (75.2.0)
Requirement already satisfied: six>=1.12.0 in
/usr/local/lib/python3.11/dist-packages (from tensorflow) (1.17.0)
Requirement already satisfied: termcolor>=1.1.0 in
/usr/local/lib/python3.11/dist-packages (from tensorflow) (3.1.0)
Requirement already satisfied: typing-extensions>=3.6.6 in
/usr/local/lib/python3.11/dist-packages (from tensorflow) (4.14.0)
Requirement already satisfied: wrapt>=1.11.0 in
/usr/local/lib/python3.11/dist-packages (from tensorflow) (1.17.2)
Requirement already satisfied: grpcio<2.0,>=1.24.3 in
/usr/local/lib/python3.11/dist-packages (from tensorflow) (1.72.1)
Collecting tensorboard~=2.19.0 (from tensorflow)
  Downloading tensorboard-2.19.0-py3-none-any.whl.metadata (1.8 kB)
Requirement already satisfied: keras>=3.5.0 in
/usr/local/lib/python3.11/dist-packages (from tensorflow) (3.8.0)
Requirement already satisfied: numpy<2.2.0,>=1.26.0 in
/usr/local/lib/python3.11/dist-packages (from tensorflow) (2.0.2)
Requirement already satisfied: h5py>=3.11.0 in
/usr/local/lib/python3.11/dist-packages (from tensorflow) (3.13.0)
Collecting ml-dtypes<1.0.0,>=0.5.1 (from tensorflow)
  Downloading ml_dtypes-0.5.1-cp311-cp311-
manylinux 2 17 x86 64.manylinux2014 x86 64.whl.metadata (21 kB)
Requirement already satisfied: tensorflow-io-gcs-filesystem>=0.23.1 in
/usr/local/lib/python3.11/dist-packages (from tensorflow) (0.37.1)
Requirement already satisfied: wheel<1.0,>=0.23.0 in
/usr/local/lib/python3.11/dist-packages (from astunparse>=1.6.0-
>tensorflow) (0.45.1)
Requirement already satisfied: rich in /usr/local/lib/python3.11/dist-
packages (from keras>=3.5.0->tensorflow) (13.9.4)
Requirement already satisfied: namex in
/usr/local/lib/python3.11/dist-packages (from keras>=3.5.0-
>tensorflow) (0.1.0)
Requirement already satisfied: optree in
```

```
/usr/local/lib/python3.11/dist-packages (from keras>=3.5.0-
>tensorflow) (0.16.0)
Requirement already satisfied: charset-normalizer<4,>=2 in
/usr/local/lib/python3.11/dist-packages (from requests<3,>=2.21.0-
>tensorflow) (3.4.2)
Requirement already satisfied: idna<4,>=2.5 in
/usr/local/lib/python3.11/dist-packages (from requests<3,>=2.21.0-
>tensorflow) (3.10)
Requirement already satisfied: urllib3<3,>=1.21.1 in
/usr/local/lib/python3.11/dist-packages (from requests<3,>=2.21.0-
>tensorflow) (2.4.0)
Requirement already satisfied: certifi>=2017.4.17 in
/usr/local/lib/python3.11/dist-packages (from requests<3,>=2.21.0-
>tensorflow) (2025.4.26)
Requirement already satisfied: markdown>=2.6.8 in
/usr/local/lib/python3.11/dist-packages (from tensorboard~=2.19.0-
>tensorflow) (3.8)
Requirement already satisfied: tensorboard-data-server<0.8.0,>=0.7.0
in /usr/local/lib/python3.11/dist-packages (from tensorboard~=2.19.0-
>tensorflow) (0.7.2)
Requirement already satisfied: werkzeug>=1.0.1 in
/usr/local/lib/python3.11/dist-packages (from tensorboard~=2.19.0-
>tensorflow) (3.1.3)
Requirement already satisfied: MarkupSafe>=2.1.1 in
/usr/local/lib/python3.11/dist-packages (from werkzeug>=1.0.1-
>tensorboard~=2.19.0->tensorflow) (3.0.2)
Requirement already satisfied: markdown-it-py>=2.2.0 in
/usr/local/lib/python3.11/dist-packages (from rich->keras>=3.5.0-
>tensorflow) (3.0.0)
Requirement already satisfied: pygments<3.0.0,>=2.13.0 in
/usr/local/lib/python3.11/dist-packages (from rich->keras>=3.5.0-
>tensorflow) (2.19.1)
Requirement already satisfied: mdurl~=0.1 in
/usr/local/lib/python3.11/dist-packages (from markdown-it-py>=2.2.0-
>rich->keras>=3.5.0->tensorflow) (0.1.2)
Downloading tensorflow-2.19.0-cp311-cp311-
manylinux 2 17 x86 64.manylinux2014 x86 64.whl (644.9 MB)
                                     -- 644.9/644.9 MB 858.0 kB/s eta
0:00:00
l dtypes-0.5.1-cp311-cp311-
manylinux 2 17 x86 64.manylinux2014 x86 64.whl (4.7 MB)
                                      4.7/4.7 MB 81.7 MB/s eta
0:00:00
                                      -- 5.5/5.5 MB 76.7 MB/s eta
0:00:00
l-dtypes, tensorboard, tensorflow
  Attempting uninstall: ml-dtypes
    Found existing installation: ml-dtypes 0.4.1
    Uninstalling ml-dtypes-0.4.1:
```

```
Successfully uninstalled ml-dtypes-0.4.1
  Attempting uninstall: tensorboard
    Found existing installation: tensorboard 2.18.0
    Uninstalling tensorboard-2.18.0:
      Successfully uninstalled tensorboard-2.18.0
 Attempting uninstall: tensorflow
    Found existing installation: tensorflow 2.18.0
    Uninstalling tensorflow-2.18.0:
      Successfully uninstalled tensorflow-2.18.0
ERROR: pip's dependency resolver does not currently take into account
all the packages that are installed. This behaviour is the source of
the following dependency conflicts.
tensorflow-decision-forests 1.11.0 requires tensorflow==2.18.0, but
you have tensorflow 2.19.0 which is incompatible.
tensorflow-text 2.18.1 requires tensorflow<2.19,>=2.18.0, but you have
tensorflow 2.19.0 which is incompatible.
tf-keras 2.18.0 requires tensorflow<2.19,>=2.18, but you have
tensorflow 2.19.0 which is incompatible.
Successfully installed ml-dtypes-0.5.1 tensorboard-2.19.0 tensorflow-
2.19.0
import tensorflow as tf
from tensorflow.keras.models import Model
from tensorflow.keras.layers import Input, Dense
input_dim = X_scaled.shape[1]
input layer = Input(shape=(input dim,))
encoded = Dense(16, activation='relu')(input layer)
encoded = Dense(8, activation='relu')(encoded)
decoded = Dense(16, activation='relu')(encoded)
output layer = Dense(input dim, activation='linear')(decoded)
autoencoder = Model(inputs=input layer, outputs=output layer)
autoencoder.compile(optimizer='adam', loss='mse')
autoencoder.fit(X scaled, X scaled, epochs=10, batch size=256,
shuffle=True, validation split=0.2)
reconstructions = autoencoder.predict(X scaled)
mse = tf.keras.losses.mse(X_scaled, reconstructions).numpy()
threshold = mse.mean() + 3 * mse.std()
y pred auto = [1 if e > threshold else 0 for e in mse]
print("
    Autoencoder Results:")
print(confusion matrix(y, y pred auto))
print(classification report(y, y pred auto))
```

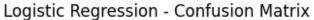
```
Epoch 1/10
                            - 4s 3ms/step - loss: 0.9119 - val loss:
891/891 -
0.7111
Epoch 2/10
891/891 -
                            - 2s 3ms/step - loss: 0.6220 - val loss:
0.6301
Epoch 3/10
891/891 -
                             - 3s 3ms/step - loss: 0.5770 - val loss:
0.5951
Epoch 4/10
891/891 —
                             4s 2ms/step - loss: 0.5423 - val loss:
0.5685
Epoch 5/10
                            - 3s 2ms/step - loss: 0.5248 - val loss:
891/891 -
0.5583
Epoch 6/10
891/891 -
                             - 3s 2ms/step - loss: 0.5129 - val loss:
0.5446
Epoch 7/10
                             - 3s 3ms/step - loss: 0.4974 - val loss:
891/891 -
0.5390
Epoch 8/10
                            - 5s 3ms/step - loss: 0.4919 - val loss:
891/891 -
0.5359
Epoch 9/10
891/891 -
                            - 2s 2ms/step - loss: 0.4845 - val_loss:
0.5332
Epoch 10/10
                             2s 3ms/step - loss: 0.4830 - val loss:
891/891 -
0.5312
8901/8901 -
                              10s 1ms/step

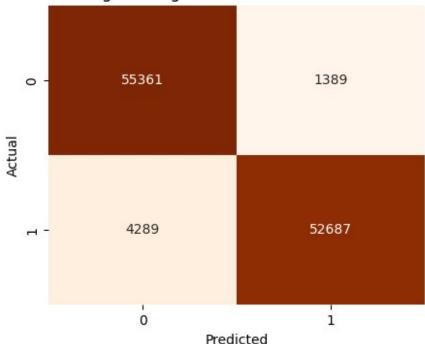
  □ Autoencoder Results:

[[283675]
            6401
     469
             23]]
                            recall f1-score
              precision
                                                support
           0
                    1.00
                              1.00
                                        1.00
                                                 284315
           1
                    0.03
                              0.05
                                        0.04
                                                    492
                                        1.00
                                                 284807
    accuracy
                    0.52
                              0.52
                                        0.52
                                                 284807
   macro avq
                                        1.00
weighted avg
                    1.00
                              1.00
                                                 284807
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.metrics import confusion matrix
def plot confusion(model_name, y_true, y_pred):
    cm = confusion matrix(y true, y pred)
```

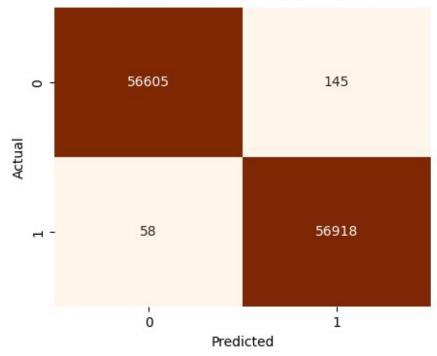
```
plt.figure(figsize=(5, 4))
    sns.heatmap(cm, annot=True, fmt='d', cmap='0ranges', cbar=False)
    plt.title(f"{model_name} - Confusion Matrix")
    plt.xlabel("Predicted")
    plt.ylabel("Actual")
    plt.show()

plot_confusion("Logistic Regression", y_test, lr_pred)
plot_confusion("Decision Tree", y_test, dt_pred)
plot_confusion("Neural Network", y_test, mlp_pred)
```

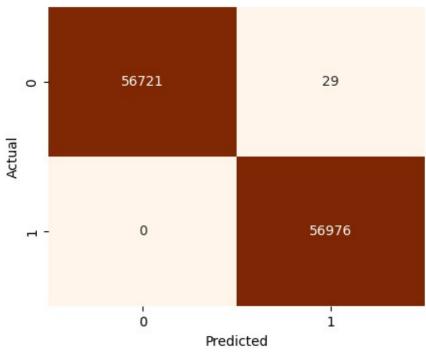








Neural Network - Confusion Matrix



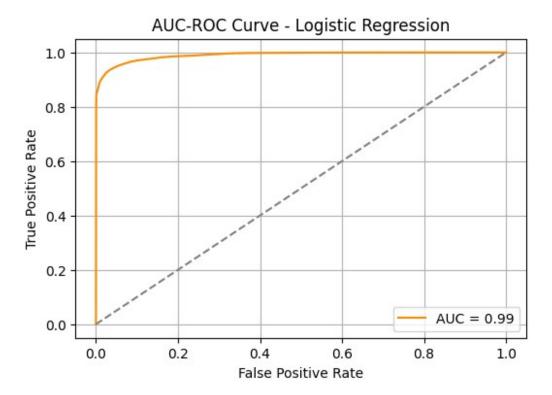
from sklearn.metrics import roc_curve, roc_auc_score
import matplotlib.pyplot as plt

```
y_prob = lr.predict_proba(X_test)[:, 1]

auc_score = roc_auc_score(y_test, y_prob)

fpr, tpr, thresholds = roc_curve(y_test, y_prob)

plt.figure(figsize=(6, 4))
plt.plot(fpr, tpr, label=f'AUC = {auc_score:.2f}', color='darkorange')
plt.plot([0, 1], [0, 1], linestyle='--', color='gray')
plt.title('AUC-ROC Curve - Logistic Regression')
plt.xlabel('False Positive Rate')
plt.ylabel('True Positive Rate')
plt.legend()
plt.grid(True)
plt.show()
```



```
from sklearn.metrics import accuracy_score
import seaborn as sns

models = ['Logistic Regression', 'Decision Tree', 'Neural Network']
accuracies = [
    accuracy_score(y_test, lr_pred),
    accuracy_score(y_test, dt_pred),
```

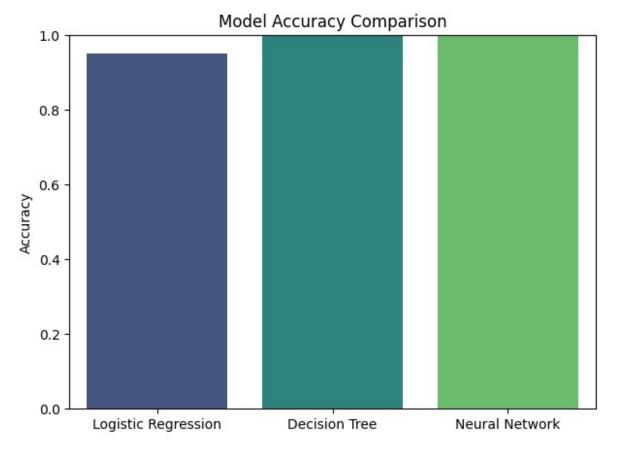
```
accuracy_score(y_test, mlp_pred)
]

plt.figure(figsize=(7, 5))
sns.barplot(x=models, y=accuracies, palette="viridis")
plt.title("Model Accuracy Comparison")
plt.ylabel("Accuracy")
plt.ylim(0, 1)
plt.show()

<ipython-input-24-2395171501>:12: FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `legend=False` for the same effect.

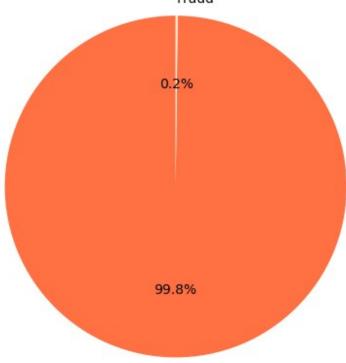
sns.barplot(x=models, y=accuracies, palette="viridis")
```



```
import matplotlib.pyplot as plt
labels = ['Normal', 'Fraud']
sizes = y.value_counts()
```

```
plt.figure(figsize=(5, 5))
plt.pie(sizes, labels=labels, autopct='%1.1f%%', startangle=90,
colors=['#ff7043' ,'#fff3e0'])
plt.title("Original Class Distribution")
plt.axis('equal')
plt.show()
```





Normal