

```
# Import necessary libraries
import numpy as np
import pandas as pd
import os
import matplotlib.pyplot as plt
from scipy.signal import butter, lfilter
from sklearn.model_selection import train_test_split
from tensorflow.keras.models import Model
from tensorflow.keras.layers import Input, Conv2D, MaxPooling2D, Flatten, Dense, Dropout
from tensorflow.keras.applications import VGG16
from tensorflow.keras.preprocessing.image import ImageDataGenerator
from tensorflow.keras.callbacks import EarlyStopping, ModelCheckpoint
```

```
from google.colab import drive
drive.mount('/content/drive')
```

↻ Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive", force\_remount=True).

```
import pandas as pd
from sklearn.preprocessing import StandardScaler, LabelEncoder
import matplotlib.pyplot as plt
import seaborn as sns
```

```
# Adjust this path to your file's actual location
file_path = '/content/drive/MyDrive/datsaforones/S1_E2_A1.csv'
data = pd.read_csv(file_path)
```

```
print("First few rows of the dataset:")
print(data.head())
```

```
print("\nDataset information:")
print(data.info())
```



```

67 18.1 1281644 non-null float64
68 19.1 1281644 non-null float64
69 20.1 1281644 non-null float64
70 21.1 1281644 non-null float64
71 0.4 1281644 non-null float64
72 1.3 1281644 non-null float64
73 0.5 1 non-null float64
74 0.6 1 non-null float64
75 0.7 1281644 non-null int64
76 0.8 1281644 non-null int64
77 0.9 1281644 non-null int64

```

dtypes: float64(74), int64(4)

memory usage: 762.7 MB

None

data.info()

```

22 10.1 1281644 non-null float64
23 11.1 1281644 non-null float64
24 12 1281644 non-null float64
25 13 1281644 non-null float64
26 14 1281644 non-null float64
27 15 1281644 non-null float64
28 16 1281644 non-null float64
29 17 1281644 non-null float64
30 18 1281644 non-null float64
31 19 1281644 non-null float64
32 20 1281644 non-null float64
33 21 1281644 non-null float64
34 22 1281644 non-null float64
35 23 1281644 non-null float64
36 24 1281644 non-null float64
37 25 1281644 non-null float64
38 26 1281644 non-null float64
39 27 1281644 non-null float64
40 28 1281644 non-null float64
41 29 1281644 non-null float64
42 30 1281644 non-null float64
43 31 1281644 non-null float64
44 32 1281644 non-null float64
45 33 1281644 non-null float64
46 34 1281644 non-null float64
47 35 1281644 non-null float64
48 0.2 1281644 non-null int64
49 0.3 1281644 non-null float64
50 1.2 1281644 non-null float64
51 2.2 1281644 non-null float64
52 3.2 1281644 non-null float64
53 4.2 1281644 non-null float64
54 5.2 1281644 non-null float64
55 6.2 1281644 non-null float64
56 7.2 1281644 non-null float64
57 8.2 1281644 non-null float64
58 9.2 1281644 non-null float64
59 10.2 1281644 non-null float64
60 11.2 1281644 non-null float64
61 12.1 1281644 non-null float64
62 13.1 1281644 non-null float64
63 14.1 1281644 non-null float64
64 15.1 1281644 non-null float64
65 16.1 1281644 non-null float64
66 17.1 1281644 non-null float64
67 18.1 1281644 non-null float64
68 19.1 1281644 non-null float64
69 20.1 1281644 non-null float64
70 21.1 1281644 non-null float64
71 0.4 1281644 non-null float64
72 1.3 1281644 non-null float64
73 0.5 1 non-null float64
74 0.6 1 non-null float64
75 0.7 1281644 non-null int64
76 0.8 1281644 non-null int64
77 0.9 1281644 non-null int64

```

dtypes: float64(74), int64(4)

memory usage: 762.7 MB

```
# Preprocess the data by removing all NaN values
data_cleaned = data.fillna(method='ffill')

print("\nFirst few rows of the cleaned dataset:")
print(data_cleaned.head())

print("\nCleaned dataset information:")
print(data_cleaned.info())
```

```
23  11.1    1281644 non-null  float64
24   12     1281644 non-null  float64
25   13     1281644 non-null  float64
26   14     1281644 non-null  float64
27   15     1281644 non-null  float64
28   16     1281644 non-null  float64
29   17     1281644 non-null  float64
30   18     1281644 non-null  float64
31   19     1281644 non-null  float64
32   20     1281644 non-null  float64
33   21     1281644 non-null  float64
34   22     1281644 non-null  float64
35   23     1281644 non-null  float64
36   24     1281644 non-null  float64
37   25     1281644 non-null  float64
38   26     1281644 non-null  float64
39   27     1281644 non-null  float64
40   28     1281644 non-null  float64
41   29     1281644 non-null  float64
42   30     1281644 non-null  float64
43   31     1281644 non-null  float64
44   32     1281644 non-null  float64
45   33     1281644 non-null  float64
46   34     1281644 non-null  float64
47   35     1281644 non-null  float64
48   0.2     1281644 non-null  int64
49   0.3     1281644 non-null  float64
50   1.2     1281644 non-null  float64
51   2.2     1281644 non-null  float64
52   3.2     1281644 non-null  float64
53   4.2     1281644 non-null  float64
54   5.2     1281644 non-null  float64
55   6.2     1281644 non-null  float64
56   7.2     1281644 non-null  float64
57   8.2     1281644 non-null  float64
58   9.2     1281644 non-null  float64
59  10.2     1281644 non-null  float64
60  11.2     1281644 non-null  float64
61  12.1     1281644 non-null  float64
62  13.1     1281644 non-null  float64
63  14.1     1281644 non-null  float64
64  15.1     1281644 non-null  float64
65  16.1     1281644 non-null  float64
66  17.1     1281644 non-null  float64
67  18.1     1281644 non-null  float64
68  19.1     1281644 non-null  float64
69  20.1     1281644 non-null  float64
70  21.1     1281644 non-null  float64
71   0.4     1281644 non-null  float64
72   1.3     1281644 non-null  float64
73   0.5     1281644 non-null  float64
74   0.6     1281644 non-null  float64
75   0.7     1281644 non-null  int64
76   0.8     1281644 non-null  int64
77   0.9     1281644 non-null  int64
dtypes: float64(74), int64(4)
memory usage: 762.7 MB
None
```

```
# Check if the cleaned dataset is empty
if data_cleaned.empty:
    raise ValueError("The cleaned dataset is empty. Please check your data and preprocessing steps.")

# Encode categorical columns if any exist
if 'Category' in data_cleaned.columns:
    label_encoder = LabelEncoder()
    data_cleaned['Category'] = label_encoder.fit_transform(data_cleaned['Category'])
    print("\nCategorical columns encoded.")

# Select numerical columns for scaling
numerical_columns = data_cleaned.select_dtypes(include=['float64', 'int64']).columns
scaler = StandardScaler()

# Check if there are any numerical columns left to scale
if len(numerical_columns) > 0:
    data_cleaned[numerical_columns] = scaler.fit_transform(data_cleaned[numerical_columns])
    print("\nNumerical columns normalized or scaled.")
else:
    print("No numerical columns to scale.")
```



Numerical columns normalized or scaled.

```
from sklearn.preprocessing import StandardScaler, LabelEncoder
from sklearn.model_selection import train_test_split

if 'Category' in data_cleaned.columns:
    label_encoder = LabelEncoder()
    data_cleaned['Category'] = label_encoder.fit_transform(data_cleaned['Category'])

numerical_columns = data_cleaned.select_dtypes(include=['float64', 'int64']).columns
scaler = StandardScaler()
data_cleaned[numerical_columns] = scaler.fit_transform(data_cleaned[numerical_columns])

target_column = '1.1'

X = data_cleaned.drop(columns=[target_column])
y = data_cleaned[target_column]

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

import tensorflow as tf
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense, Dropout

model = Sequential()
# Input layer
model.add(Dense(units=256, activation='relu', input_shape=(77,1)))
model.add(Dropout(0.3))
model.add(Dense(units=128, activation='relu'))
model.add(Dropout(0.3))
model.add(Dense(units=64, activation='relu'))
model.add(Dropout(0.3))
model.add(Dense(units=1, activation='relu'))
model.compile(optimizer='adam', loss='mean_squared_error', metrics=['accuracy'])

import tensorflow as tf
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense, Dropout

# Assuming your data has 77 features
input_shape = (77,)

# Building the neural network model
model = Sequential()

# Input layer
model.add(Dense(units=256, activation='relu', input_shape=input_shape))
```

```
model.add(Dense(units=256, activation='relu', input_shape=input_shape))
model.add(Dropout(0.3))

# Hidden layers
model.add(Dense(units=128, activation='relu'))
model.add(Dropout(0.3))

model.add(Dense(units=64, activation='relu'))
model.add(Dropout(0.3))

# Output layer
model.add(Dense(units=1, activation='sigmoid')) # Assuming binary classification

# Compile the model
model.compile(optimizer='adam', loss='binary_crossentropy', metrics=['accuracy'])

# Summary of the model
model.summary()

# Train the model with adjusted parameters
history = model.fit(X_train, y_train, epochs=10, batch_size=32, validation_split=0.2)

# Evaluate the model
loss, accuracy = model.evaluate(X_test, y_test)
print(f"Model Test Accuracy: {accuracy}")

# Plot the training history
import matplotlib.pyplot as plt

plt.figure(figsize=(12, 6))
plt.plot(history.history['accuracy'], label='Training Accuracy')
plt.plot(history.history['val_accuracy'], label='Validation Accuracy')
plt.title('Model Accuracy')
plt.xlabel('Epochs')
plt.ylabel('Accuracy')
plt.legend()
plt.show()

plt.figure(figsize=(12, 6))
plt.plot(history.history['loss'], label='Training Loss')
plt.plot(history.history['val_loss'], label='Validation Loss')
plt.title('Model Loss')
plt.xlabel('Epochs')
plt.ylabel('Loss')
plt.legend()
plt.show()

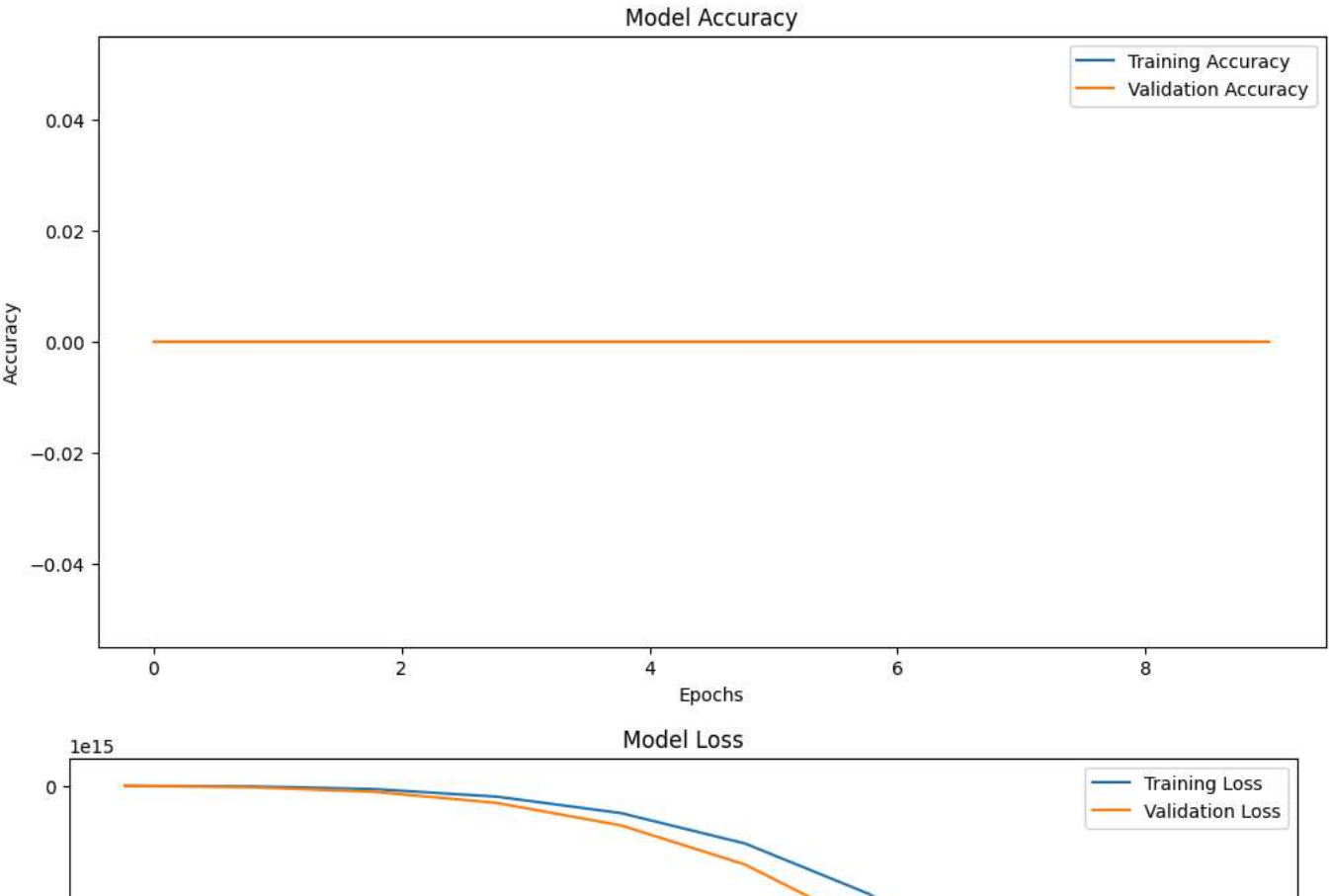
# Save the trained model for future use
model_file_path = '/content/drive/MyDrive/trained_model.h5'
model.save(model_file_path)
print(f"\nTrained model saved to {model_file_path}")
```

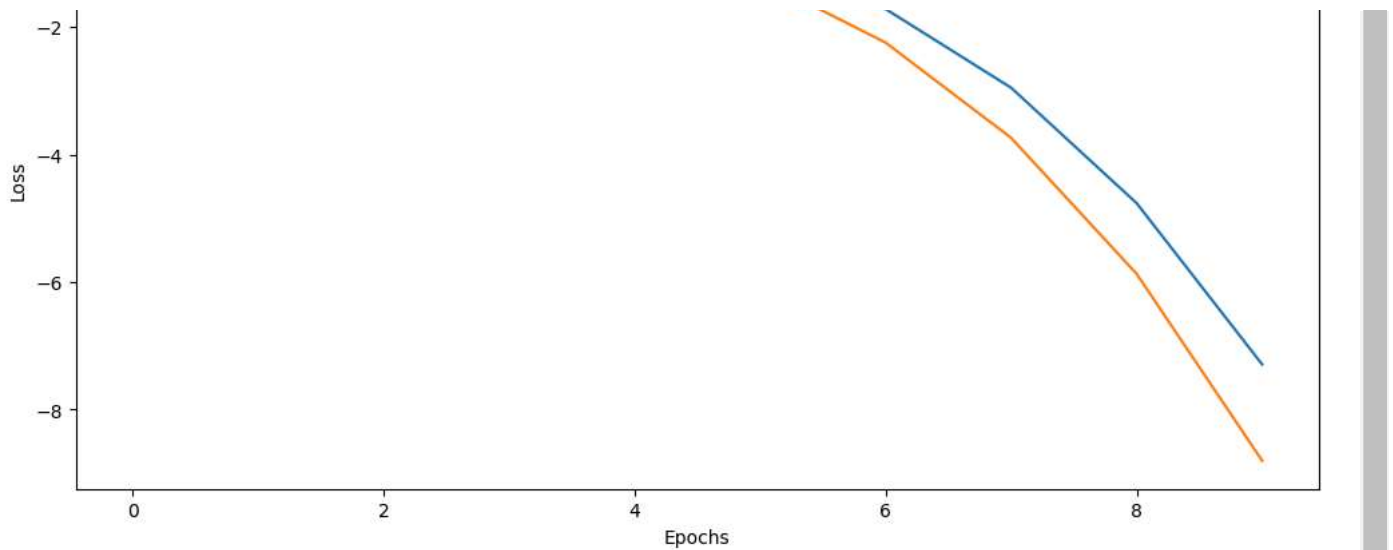
Model: "sequential\_5"

Layer (type)	Output Shape	Param #
dense_20 (Dense)	(None, 256)	19968
dropout_15 (Dropout)	(None, 256)	0
dense_21 (Dense)	(None, 128)	32896
dropout_16 (Dropout)	(None, 128)	0
dense_22 (Dense)	(None, 64)	8256
dropout_17 (Dropout)	(None, 64)	0
dense_23 (Dense)	(None, 1)	65

=====  
Total params: 61185 (239.00 KB)  
Trainable params: 61185 (239.00 KB)  
Non-trainable params: 0 (0.00 Byte)

Epoch 1/10  
25633/25633 [=====] - 108s 4ms/step - loss: -451477143552.0000 - accuracy: 0.0000e+00 - val\_loss: -195794187  
Epoch 2/10  
25633/25633 [=====] - 105s 4ms/step - loss: -9280306544640.0000 - accuracy: 0.0000e+00 - val\_loss: -21813511  
Epoch 3/10  
25633/25633 [=====] - 113s 4ms/step - loss: -52059734802432.0000 - accuracy: 0.0000e+00 - val\_loss: -9455864  
Epoch 4/10  
25633/25633 [=====] - 104s 4ms/step - loss: -173041212456960.0000 - accuracy: 0.0000e+00 - val\_loss: -273019  
Epoch 5/10  
25633/25633 [=====] - 103s 4ms/step - loss: -433633521303552.0000 - accuracy: 0.0000e+00 - val\_loss: -628362  
Epoch 6/10  
25633/25633 [=====] - 107s 4ms/step - loss: -916453947604992.0000 - accuracy: 0.0000e+00 - val\_loss: -125038  
Epoch 7/10  
25633/25633 [=====] - 103s 4ms/step - loss: -1718055637876736.0000 - accuracy: 0.0000e+00 - val\_loss: -22448  
Epoch 8/10  
25633/25633 [=====] - 104s 4ms/step - loss: -2953293869350912.0000 - accuracy: 0.0000e+00 - val\_loss: -37390  
Epoch 9/10  
25633/25633 [=====] - 104s 4ms/step - loss: -4763449443745792.0000 - accuracy: 0.0000e+00 - val\_loss: -58720  
Epoch 10/10  
25633/25633 [=====] - 107s 4ms/step - loss: -7292828161933312.0000 - accuracy: 0.0000e+00 - val\_loss: -88028  
8011/8011 [=====] - 17s 2ms/step - loss: -8817913066684416.0000 - accuracy: 0.0000e+00  
Model Test Accuracy: 0.0





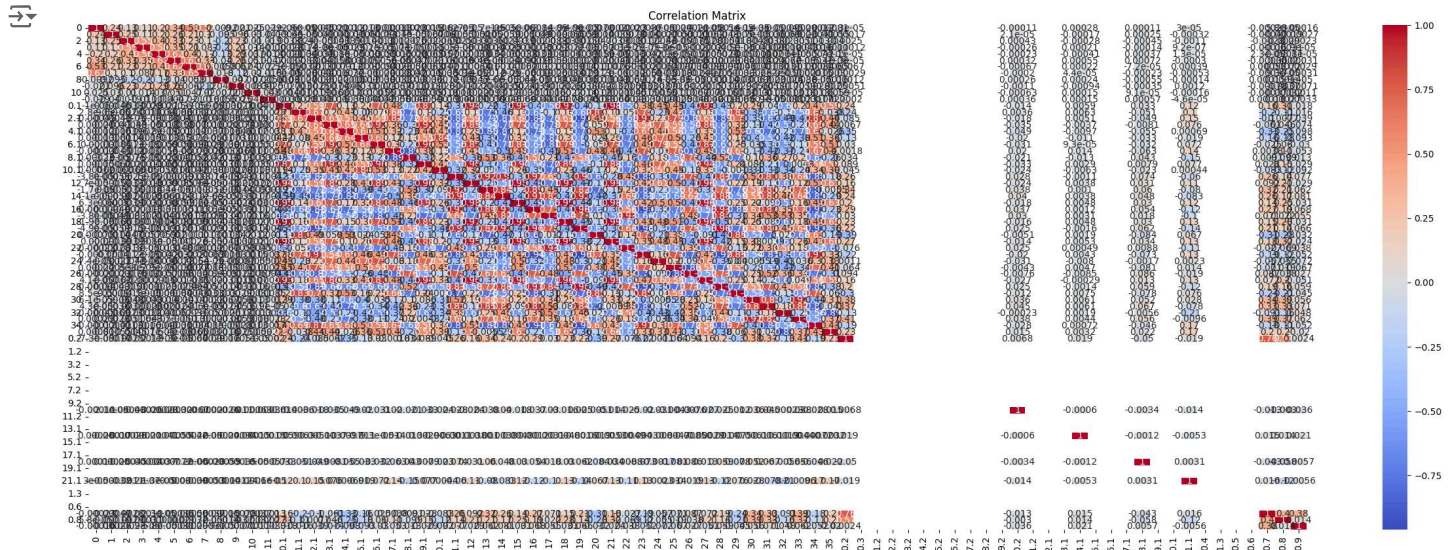
/usr/local/lib/python3.10/dist-packages/keras/src/engine/training.py:3103: UserWarning: You are saving your model as an HDF5 file via saving\_api.save\_model(

Trained model saved to /content/drive/MyDrive/trained\_model.h5

Start coding or [generate](#) with AI.

```
# Plotting correlation matrix and pairplot
plt.figure(figsize=(30, 10))
sns.heatmap(data_cleaned.corr(), annot=True, cmap='coolwarm')
plt.title('Correlation Matrix')
plt.show()
```

```
sns.pairplot(data_cleaned)
plt.show()
```



```
# Plot the training history
```