

```

import numpy as np
import pandas as pd
import tensorflow as tf
from tensorflow.keras.preprocessing.text import Tokenizer
from tensorflow.keras.preprocessing.sequence import pad_sequences
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Embedding, Conv1D, GlobalMaxPooling1D, Dense, Dropout
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import LabelEncoder

```

```
# Replace 'your_file.csv' with the actual filename you uploaded
```

```
df = pd.read_csv("Review_db.csv") # Read the CSV file into a DataFrame
```

```
# Display the first 5 rows
print(df.head())
```



	City	Place \
0	Aamby Valley City	19 Degree North
1	Aamby Valley City	19 Degree North
2	Aamby Valley City	19 Degree North
3	Aamby Valley City	19 Degree North
4	Aamby Valley City	19 Degree North

	Review	Rating	Name	Date \
0	aamby valley beautiful place clear blue skies ...	5	Anonymous	NaN
1	executed obt akshay thanx team thoroughly enjo...	4	Anonymous	NaN
2	awesome experience atv tracts obstacles mainta...	5	Anonymous	NaN
3	visited aamby valley yesterday short excursion...	4	Anonymous	NaN
4	far mumbai place finest adventure places visit...	5	Anonymous	NaN

	Raw_Review
0	Aamby valley is a beautiful place with its cle...
1	Very well executed obt by Akshay.... Thanx as ...
2	Awesome experience at the ATV\nTracts and obst...
3	we visited the Aamby Valley yesterday for shor...
4	Not far from Mumbai, this place is one of the ...

```
# Define a function to map ratings to sentiment categories
```

```
def map_rating_to_sentiment(rating):
    if rating >= 4:
        return "Positive"
    elif rating == 3:
        return "Neutral"
    else:
        return "Negative"
```

```
# Apply the function to create the sentiment column
```

```
df['sentiment'] = df['Rating'].apply(map_rating_to_sentiment)
```

```
# Drop the rating column (optional)
```

```
df.drop(columns=['Rating'], inplace=True)
```

```
# Display first few rows to verify
print(df.head())
```

```

City      Place \
0  Aamby Valley City  19 Degree North
1  Aamby Valley City  19 Degree North
2  Aamby Valley City  19 Degree North
3  Aamby Valley City  19 Degree North
4  Aamby Valley City  19 Degree North

Review      Name  Date \
0  aamby valley beautiful place clear blue skies ...  Anonymous  NaN
1  executed obt akshay thanx team thoroughly enjo...  Anonymous  NaN
2  awesome experience atv tracts obstacles mainta...  Anonymous  NaN
3  visited aamby valley yesterday short excursion...  Anonymous  NaN
4  far mumbai place finest adventure places visit...  Anonymous  NaN

Raw_Review  sentiment
0  Aamby valley is a beautiful place with its cle...  Positive
1  Very well executed obt by Akshay.... Thanx as ...  Positive
2  Awesome experience at the ATV\nTracts and obst...  Positive
3  we visited the Aamby Valley yesterday for shor...  Positive
4  Not far from Mumbai, this place is one of the ...  Positive

```

```
from sklearn.preprocessing import LabelEncoder
```

```
label_encoder = LabelEncoder()
df['sentiment'] = label_encoder.fit_transform(df['sentiment'])
```

```
# Show mapping of labels
```

```
label_mapping = dict(zip(label_encoder.classes_, label_encoder.transform(label_encoder.cl
print(label_mapping) # Example Output: {'Negative': 0, 'Neutral': 1, 'Positive': 2}
```

```
{'Negative': np.int64(0), 'Neutral': np.int64(1), 'Positive': np.int64(2)}
```

```
# Parameters
```

```
max_words = 10000 # Consider top 10,000 words
max_len = 100 # Maximum words per review
```

```
# Tokenization
```

```
tokenizer = Tokenizer(num_words=max_words, oov_token="<OOV>")
tokenizer.fit_on_texts(df['Review'])
```

```
# Convert text to sequences
```

```
sequences = tokenizer.texts_to_sequences(df['Review'])
padded_sequences = pad_sequences(sequences, maxlen=max_len, padding='post')
```

```
# Splitting data into train and test sets
```

```
X_train, X_test, y_train, y_test = train_test_split(padded_sequences, df['sentiment'], te
```

```

model = Sequential([
    Embedding(input_dim=max_words, output_dim=128, input_length=max_len),
    Conv1D(filters=128, kernel_size=5, activation='relu'),
    GlobalMaxPooling1D(),
    Dense(64, activation='relu'),
    Dropout(0.5),
    Dense(3, activation='softmax') # 3 output classes (Negative, Neutral, Positive)
])


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

# Summary of the model
model.summary()

```

→ /usr/local/lib/python3.11/dist-packages/keras/src/layers/core/embedding.py:90: UserWarning:
 warnings.warn(
 Model: "sequential"

Layer (type)	Output Shape	Param
embedding (Embedding)	?	0 (unbuilt)
conv1d (Conv1D)	?	0 (unbuilt)
global_max_pooling1d (GlobalMaxPooling1D)	?	
dense (Dense)	?	0 (unbuilt)
dropout (Dropout)	?	
dense_1 (Dense)	?	0 (unbuilt)

◀  ▶
Total params: 0 (0.00 B)
Trainable params: 0 (0.00 B)
Non-trainable params: 0 (0.00 B)
 ▶

```


epochs = 1
batch_size = 32

```

```

history = model.fit(X_train, y_train, validation_data=(X_test, y_test), epochs=epochs, batch_size=batch_size)


```

→ 37062/37062  2116s 57ms/step - accuracy: 0.9657 - loss: 0.1015 -

```

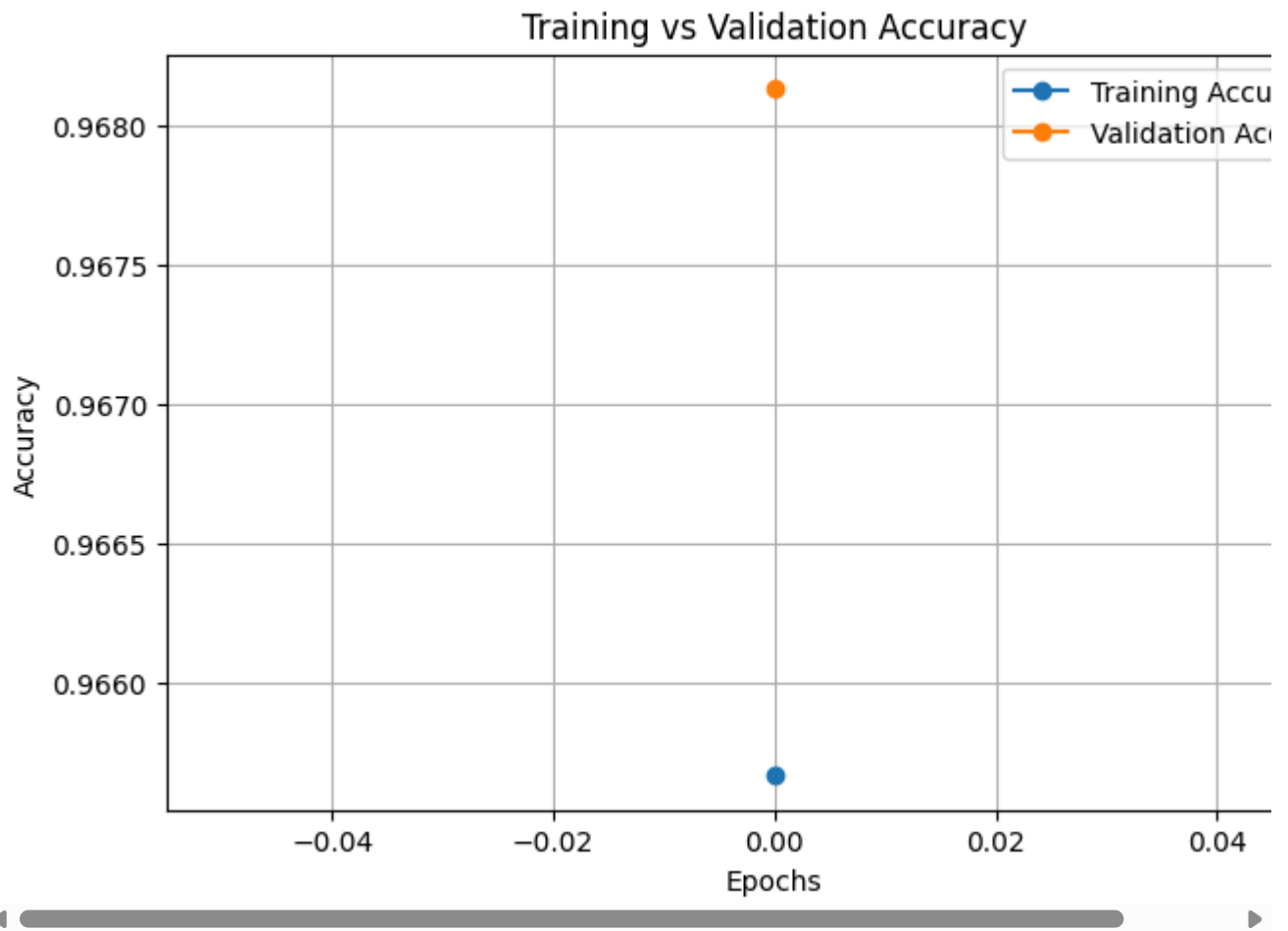
test_loss, test_accuracy = model.evaluate(X_test, y_test)
print(f"Test Accuracy: {test_accuracy:.4f}")

```

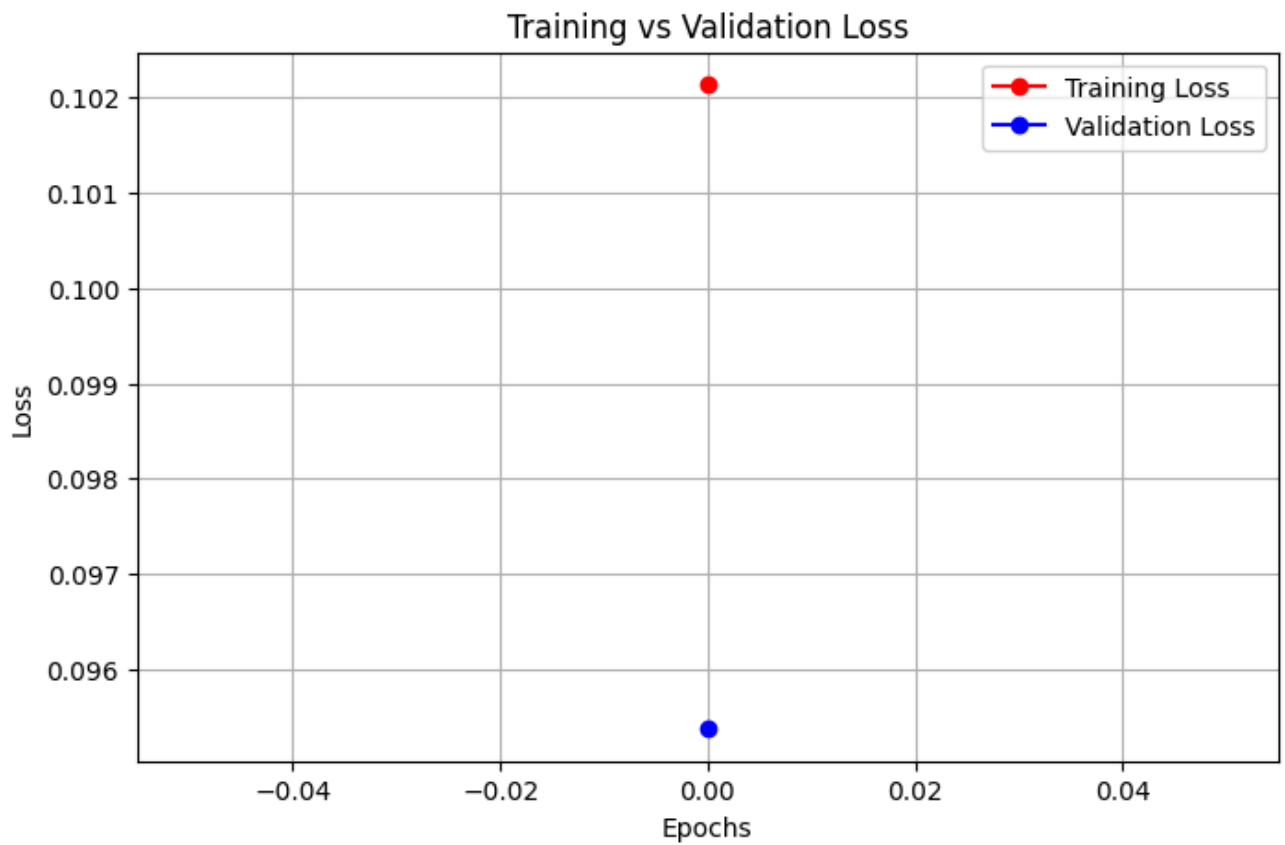
→ 9266/9266  128s 14ms/step - accuracy: 0.9679 - loss: 0.0957
 Test Accuracy: 0.9681

```
import matplotlib.pyplot as plt
```

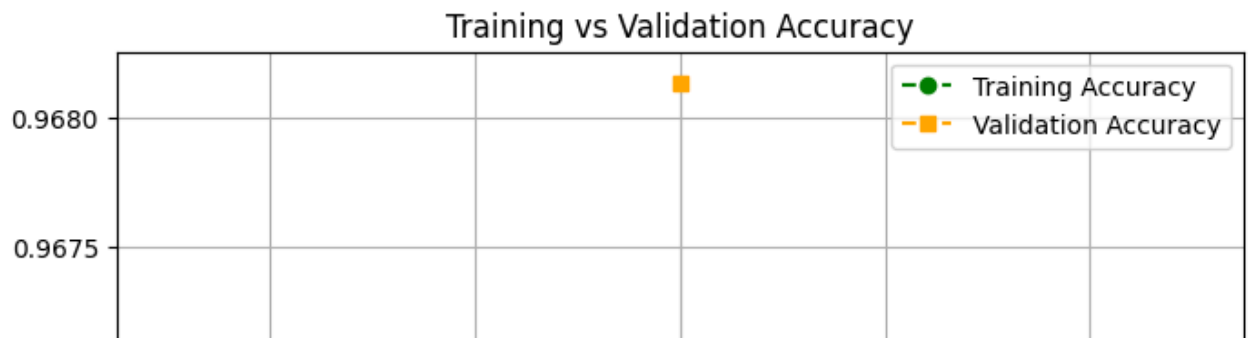
```
plt.figure(figsize=(8, 5))
plt.plot(history.history['accuracy'], label='Training Accuracy', marker='o')
plt.plot(history.history['val_accuracy'], label='Validation Accuracy', marker='o')
plt.xlabel('Epochs')
plt.ylabel('Accuracy')
plt.title('Training vs Validation Accuracy')
plt.legend()
plt.grid(True)
plt.show()
```



```
plt.figure(figsize=(8, 5))
plt.plot(history.history['loss'], label='Training Loss', marker='o', color='red')
plt.plot(history.history['val_loss'], label='Validation Loss', marker='o', color='blue')
plt.xlabel('Epochs')
plt.ylabel('Loss')
plt.title('Training vs Validation Loss')
plt.legend()
plt.grid(True)
plt.show()
```



```
plt.figure(figsize=(8, 5))
plt.plot(history.history['accuracy'], label='Training Accuracy', linestyle='dashed', marker='o')
plt.plot(history.history['val_accuracy'], label='Validation Accuracy', linestyle='dashed', marker='o')
plt.fill_between(range(len(history.history['accuracy'])), history.history['accuracy'], history.history['val_accuracy'], color='blue')
plt.xlabel('Epochs')
plt.ylabel('Accuracy')
plt.title('Training vs Validation Accuracy')
plt.legend()
plt.grid(True)
plt.show()
```



```
def predict_review(review):
    sequence = tokenizer.texts_to_sequences([review])
    padded_sequence = pad_sequences(sequence, maxlen=max_len, padding='post')
    prediction = model.predict(padded_sequence)
    sentiment = np.argmax(prediction)

    sentiment_label = {v: k for k, v in label_mapping.items()} # Reverse mapping
    return sentiment_label[sentiment]

# Example review prediction
new_review = "The place was amazing, I had a great experience!"
print("Predicted Sentiment:", predict_review(new_review))
```



1/1 ————— **0s** 134ms/step
 Predicted Sentiment: Positive