GRU, BiGRU, CNN for multiclassification

March 11, 2025

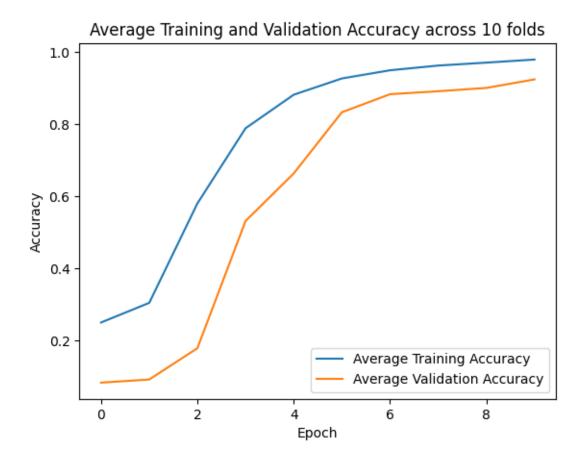
```
[1]: #GRU Model
     import pandas as pd
     import numpy as np
     import re
     from nltk.stem import WordNetLemmatizer
     from sklearn.model_selection import KFold
     from tensorflow.keras.preprocessing.text import Tokenizer # Updated import
     from tensorflow.keras.preprocessing.sequence import pad_sequences # Updatedu
      \rightarrow import
     from tensorflow.keras.models import Sequential # Updated import
     from tensorflow.keras.layers import Embedding, GRU, Dropout, Dense # Updated_
      \hookrightarrow import
     from tensorflow.keras.optimizers import Adam # Updated import
     from sklearn.metrics import confusion_matrix, classification_report
     import matplotlib.pyplot as plt
     import seaborn as sns
     from imblearn.over_sampling import RandomOverSampler
     # 1. Load and Preprocess the dataset
     #df = pd.read_csv('Amazon_Dataset_LD.csv', encoding='latin1')
     df = pd.read_csv('accessibilityissues_multi.csv', encoding='latin1')
     print(df.head())
     print(df.head())
     print(df.head())
     lemmatizer = WordNetLemmatizer()
     def clean_text(text):
         text = text.lower()
         text = re.sub(r'[^\w\s]', '', text)
         text = re.sub(r'\d+', '', text)
         tokens = text.split()
```

```
tokens = [lemmatizer.lemmatize(token) for token in tokens]
   return ' '.join(tokens)
df['Review'] = df['Review'].apply(clean_text)
tokenizer = Tokenizer()
tokenizer.fit on texts(df['Review'])
X = tokenizer.texts_to_sequences(df['Review'])
vocab_size = len(tokenizer.word_index) + 1
maxlen = 100
X = pad_sequences(X, padding='post', maxlen=maxlen)
y_dict = {'Navigation and Interaction Problems (NAV)': 0, 'Input and Control∪
 →Issues (INPUT)': 1, 'Compatibility with Assistive Technologies (CAT)': 2, □
→'UI Accessibility Issues (UI)': 3, 'Audio and Visual Accessibility issues⊔
y = df['Assessability Issue Type'].map(y_dict)
y = pd.get_dummies(df['Assessability Issue Type']).values
# 2. Oversample to balance classes
oversampler = RandomOverSampler(random_state=42)
X_resampled, y_resampled = oversampler.fit_resample(X, y)
# 3. Define the GRU model
def create_model():
   model = Sequential()
   model.add(Embedding(input_dim=vocab_size, output_dim=100,__
 →input_length=maxlen))
   model.add(GRU(64))
   model.add(Dropout(0.2))
   model.add(Dense(5, activation='softmax'))
   model.compile(optimizer=Adam(learning_rate=0.001),__
 ⇔loss='categorical_crossentropy', metrics=['accuracy'])
   return model
# 4. K-Fold Cross-Validation
n_folds = 5
kfold = KFold(n_splits=n_folds, shuffle=True)
fold no = 1
acc_per_fold = []
# Lists to store average accuracies across folds
avg_train_acc = []
avg_val_acc = []
for train, test in kfold.split(X_resampled, y_resampled):
   model = create_model()
```

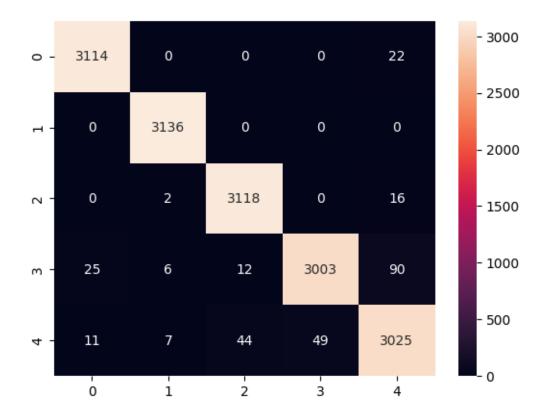
```
history = model.fit(X_resampled[train], y_resampled[train],__
  ⇒validation_split=0.1, epochs=10, batch_size=32, verbose=0)
    avg_train_acc.append(history.history['accuracy'])
    avg_val_acc.append(history.history['val_accuracy'])
    scores = model.evaluate(X_resampled[test], y_resampled[test], verbose=0)
    acc_per_fold.append(scores[1] * 100)
    fold_no += 1
# 5. Performance Metrics
print(f'> Average Accuracy across {n_folds}-folds: {np.mean(acc_per_fold):.2f}%__
 \rightarrow (+/- {np.std(acc_per_fold):.2f}%)')
# Average Training and Validation Accuracy for 5 folds
plt.plot(np.mean(avg_train_acc, axis=0), label='Average Training Accuracy')
plt.plot(np.mean(avg_val_acc, axis=0), label='Average Validation Accuracy')
plt.title('Average Training and Validation Accuracy across 5 folds')
plt.xlabel('Epoch')
plt.ylabel('Accuracy')
plt.legend()
plt.show()
# Confusion Matrix
y_pred = model.predict(X_resampled)
y_pred_classes = np.argmax(y_pred, axis=1)
y true classes = np.argmax(y resampled, axis=1)
cm = confusion_matrix(y_true_classes, y_pred_classes)
sns.heatmap(cm, annot=True, fmt='g')
plt.show()
# Classification Report
print(classification_report(y_true_classes, y_pred_classes,__
  →target names=list(y dict.keys())))
  Unnamed: 0
                    userName
                                                                          Title
0
          NaN
                   AllykatTX
                                  LYSN allows unchecked bullying & harassment
                                        Filled with toxic and abusive comments
1
          NaN
                 starry_sara
2
          \mathtt{NaN}
                                                             Toxic communities
                     dayflan
                                                               Honestlyââ¬Â¦
3
          {\tt NaN}
                myhomiejesus
4
          NaN @junmyeonsoup Despicable, toxic community for artist and fans
                                               Review Rating \
0 When I originally joined the LYSN app I expect...
                                                          1
1 Even though the companyââ¬â¢s policy states...
2 At first, I was excited about downloading the ...
                                                          1
3 I see a lot of 5 star reviews of people saying...
```

4	I expected this app to be a good time for me t 1		
0 1 2 3 4	cleaned_Review Have assessablity when i originally joined the lysn app i expect even though the company s policy state they wi at first i wa excited about downloading the ap i see a lot of star review of people saying th i expected this app to be a good time for me t	Issue Yes Yes Yes Yes	\
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	Unnamed: 0 userName	Title	\
0	NaN AllykatTX LYSN allows unchecked bullying & hara		
1	NaN starry_sara Filled with toxic and abusive co		
2	NaN dayflan Toxic commu		
3	NaN myhomiejesus Honestly/		
4	NaN @junmyeonsoup Despicable, toxic community for artist ar	id Talls	
	Review Rating \		
0	When I originally joined the LYSN app I expect 1		
1	Even though the companyââ¬â¢s policy states 1		
2	At first, I was excited about downloading the 1		
3	I see a lot of 5 star reviews of people saying 1		
4	I expected this app to be a good time for me t 1		
	cleaned_Review Have assessablity	Issue	\
0	cleaned_Review Have assessablity when i originally joined the lysn app i expect	Issue Yes	\
0	_ •		\
	when i originally joined the lysn app i expect	Yes	\
1	when i originally joined the lysn app i expect even though the company s policy state they wi	Yes Yes	\
1 2	when i originally joined the lysn app i expect even though the company s policy state they wi at first i wa excited about downloading the ap	Yes Yes	\
1 2 3	when i originally joined the lysn app i expect even though the company s policy state they wi at first i wa excited about downloading the ap i see a lot of star review of people saying th i expected this app to be a good time for me t Assessability Issue Type	Yes Yes Yes	\
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1 2 3 4	when i originally joined the lysn app i expect even though the company s policy state they wi at first i wa excited about downloading the ap i see a lot of star review of people saying th i expected this app to be a good time for me t Assessability Issue Type Navigation and Interaction Problems (NAV) Navigation and Interaction Problems (NAV) Navigation and Interaction Problems (NAV)	Yes Yes Yes	\
1 2 3 4 0 1	when i originally joined the lysn app i expect even though the company s policy state they wi at first i wa excited about downloading the ap i see a lot of star review of people saying th i expected this app to be a good time for me t Assessability Issue Type Navigation and Interaction Problems (NAV) Navigation and Interaction Problems (NAV) Navigation and Interaction Problems (NAV) Input and Control Issues (INPUT)	Yes Yes Yes	\
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1 2 3 4 0 1 2 3 4	when i originally joined the lysn app i expect even though the company s policy state they wi at first i wa excited about downloading the ap i see a lot of star review of people saying th i expected this app to be a good time for me t Assessability Issue Type Navigation and Interaction Problems (NAV) Navigation and Interaction Problems (NAV) Navigation and Interaction Problems (NAV) Input and Control Issues (INPUT) Navigation and Interaction Problems (NAV) Unnamed: 0 userName NaN AllykatTX LYSN allows unchecked bullying & harangement to be a good time for me t Assessability Issue Type NaVigation and Interaction Problems (NAV) LYSN allows unchecked bullying & harangement to be a good time for me t	Yes Yes Yes Yes Yes Title assment omments unities Å¢â¬Â¦	

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Review Rating \
O When I originally joined the LYSN app I expect...
                                                         1
1 Even though the companyââ¬â¢s policy states...
                                                         1
2 At first, I was excited about downloading the ...
                                                         1
3 I see a lot of 5 star reviews of people saying...
                                                          1
4 I expected this app to be a good time for me t...
                                      cleaned Review Have assessablity Issue \
0 when i originally joined the lysn app i expect...
1 even though the company s policy state they wi...
                                                                        Yes
2 at first i wa excited about downloading the ap...
                                                                        Yes
3 i see a lot of star review of people saying th...
                                                                        Yes
4 i expected this app to be a good time for me t...
                                                                        Yes
                    Assessability Issue Type
O Navigation and Interaction Problems (NAV)
1 Navigation and Interaction Problems (NAV)
2 Navigation and Interaction Problems (NAV)
3
            Input and Control Issues (INPUT)
4 Navigation and Interaction Problems (NAV)
/Users/nekdilkhan/miniforge3/lib/python3.12/site-
packages/keras/src/layers/core/embedding.py:90: UserWarning: Argument
`input_length` is deprecated. Just remove it.
  warnings.warn(
```



490/490 4s 9ms/step



		precision	recall
f1-score	support		
	Navigation and Interaction Problems (NAV)	0.99	0.99
0.99	3136		
	Input and Control Issues (INPUT)	1.00	1.00
1.00	3136		
Compa	atibility with Assistive Technologies (CAT)	0.98	0.99
0.99	3136		
	UI Accessibility Issues (UI)	0.98	0.96
0.97	3136		
Audio and	d Visual Accessibility issues (AUDIOVISUAL)	0.96	0.96
0.96	3136		
	accuracy		
0.98	15680		
	macro avg	0.98	0.98
0.98	15680		
	weighted avg	0.98	0.98
0.98	15680		

```
[2]: # BiGRU Model
     import pandas as pd
     import numpy as np
     import re
     from nltk.stem import WordNetLemmatizer
     from sklearn.model_selection import KFold
     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, Bidirectional, GRU, Dropout,
      ⊶Dense
     from tensorflow.keras.optimizers import Adam
     from sklearn.metrics import confusion_matrix, classification_report
     import matplotlib.pyplot as plt
     import seaborn as sns
     from imblearn.over sampling import RandomOverSampler
     # 1. Load and Preprocess the dataset
     #df = pd.read_csv('Amazon_Dataset_LD.csv', encoding='latin1')
     df = pd.read_csv('accessibilityissues_multi.csv', encoding='latin1')
     lemmatizer = WordNetLemmatizer()
     def clean_text(text):
         text = text.lower()
         text = re.sub(r'[^\w\s]', '', text)
         text = re.sub(r'\d+', '', text)
         tokens = text.split()
         tokens = [lemmatizer.lemmatize(token) for token in tokens]
         return ' '.join(tokens)
     df['Review'] = df['Review'].apply(clean_text)
     tokenizer = Tokenizer()
     tokenizer.fit_on_texts(df['Review'])
     X = tokenizer.texts_to_sequences(df['Review'])
     vocab_size = len(tokenizer.word_index) + 1
     maxlen = 100
     X = pad_sequences(X, padding='post', maxlen=maxlen)
     y_dict = {'Navigation and Interaction Problems (NAV)': 0, 'Input and Control∪
      →Issues (INPUT)': 1, 'Compatibility with Assistive Technologies (CAT)': 2, ⊔
     →'UI Accessibility Issues (UI)': 3, 'Audio and Visual Accessibility issues
     ⇔(AUDIOVISUAL)': 4}
     y = df['Assessability Issue Type'].map(y dict)
     y = pd.get_dummies(df['Assessability Issue Type']).values
     # 2. Oversample to balance classes
     oversampler = RandomOverSampler(random_state=42)
```

```
X_resampled, y_resampled = oversampler.fit_resample(X, y)
# 3. Define the BiGRU model
def create_bigru_model():
   model = Sequential()
    model.add(Embedding(input_dim=vocab_size, output_dim=100,__
 →input_length=maxlen))
    model.add(Bidirectional(GRU(64, return_sequences=True)))
    model.add(Dropout(0.2))
    model.add(Bidirectional(GRU(64)))
    model.add(Dropout(0.2))
    model.add(Dense(5, activation='softmax'))
    model.compile(optimizer=Adam(learning_rate=0.001),__
 ⇔loss='categorical_crossentropy', metrics=['accuracy'])
    return model
# 4. K-Fold Cross-Validation
n folds = 5
kfold = KFold(n_splits=n_folds, shuffle=True)
fold_no = 1
acc_per_fold = []
# Lists to store average accuracies across folds
avg_train_acc = []
avg_val_acc = []
for train, test in kfold.split(X_resampled, y_resampled):
    model = create bigru model()
    history = model.fit(X_resampled[train], y_resampled[train],__
 →validation_split=0.1, epochs=10, batch_size=32, verbose=0)
    avg_train_acc.append(history.history['accuracy'])
    avg_val_acc.append(history.history['val_accuracy'])
    scores = model.evaluate(X_resampled[test], y_resampled[test], verbose=0)
    acc_per_fold.append(scores[1] * 100)
    fold_no += 1
# 5. Performance Metrics
print(f'> Average Accuracy across {n folds}-folds: {np.mean(acc per fold):.2f}%11
 \hookrightarrow(+/- {np.std(acc_per_fold):.2f}%)')
# Average Training and Validation Accuracy for 5 folds
plt.plot(np.mean(avg_train_acc, axis=0), label='Average Training Accuracy')
plt.plot(np.mean(avg_val_acc, axis=0), label='Average Validation Accuracy')
plt.title('Average Training and Validation Accuracy across 10 folds')
plt.xlabel('Epoch')
```

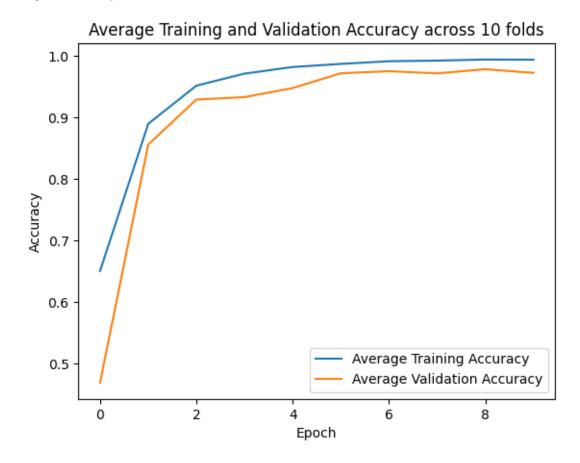
```
plt.ylabel('Accuracy')
plt.legend()
plt.show()

# Confusion Matrix
y_pred = model.predict(X_resampled)
y_pred_classes = np.argmax(y_pred, axis=1)
y_true_classes = np.argmax(y_resampled, axis=1)
cm = confusion_matrix(y_true_classes, y_pred_classes)
sns.heatmap(cm, annot=True, fmt='g')
plt.show()

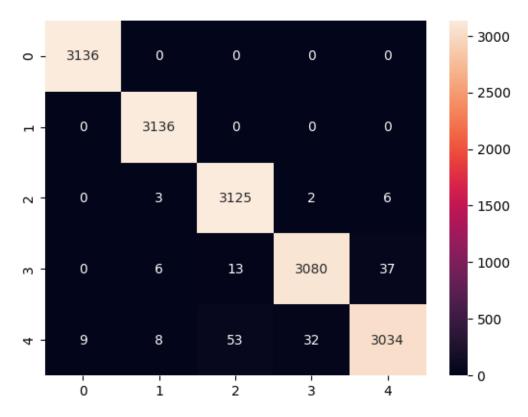
# Classification Report
print(classification_report(y_true_classes, y_pred_classes, u_dataget_names=list(y_dict.keys())))
```

/Users/nekdilkhan/miniforge3/lib/python3.12/sitepackages/keras/src/layers/core/embedding.py:90: UserWarning: Argument `input_length` is deprecated. Just remove it. warnings.warn(

> Average Accuracy across 5-folds: 95.85% (+/- 0.53%)



490/490 19s 37ms/step

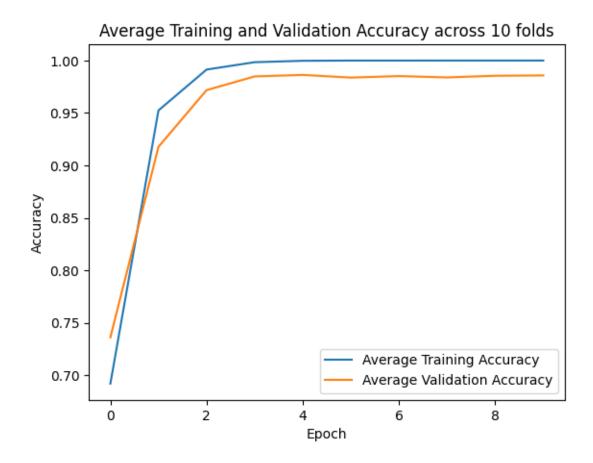


		precision	recall
f1-score	support		
	Navigation and Interaction Problems (NAV)	1.00	1.00
1.00	3136		
	Input and Control Issues (INPUT)	0.99	1.00
1.00	3136		
Compa	atibility with Assistive Technologies (CAT)	0.98	1.00
0.99	3136		
	UI Accessibility Issues (UI)	0.99	0.98
0.99	3136		
Audio and	d Visual Accessibility issues (AUDIOVISUAL)	0.99	0.97
0.98	3136		
	accuracy		
0.99	15680		
	macro avg	0.99	0.99
0.99	15680		
	weighted avg	0.99	0.99

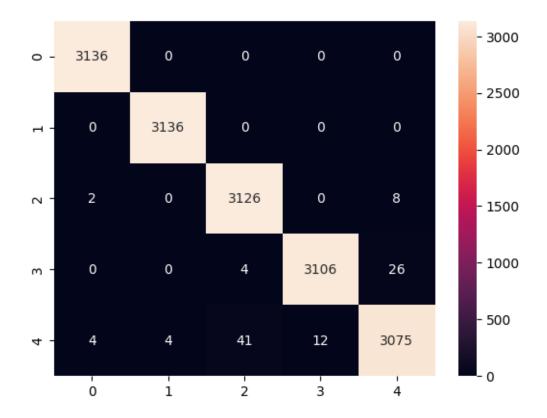
0.99 15680

```
[3]: #CNN Dataset , Oversampling ,
     import pandas as pd
     import numpy as np
     import re
     from nltk.stem import WordNetLemmatizer
     from sklearn.model selection import KFold
     from tensorflow.keras.preprocessing.text import Tokenizer # Updated import
     from tensorflow.keras.preprocessing.sequence import pad_sequences # Updated_
      \hookrightarrow import
     from tensorflow.keras.models import Sequential # Updated import
     from tensorflow.keras.layers import Embedding, Conv1D, MaxPooling1D, Flatten, U
      →Dropout, Dense # Updated import
     from tensorflow.keras.optimizers import Adam # Updated import
     from sklearn.metrics import confusion_matrix, classification_report
     import matplotlib.pyplot as plt
     import seaborn as sns
     from imblearn.over_sampling import RandomOverSampler
     # 3. Define the CNN model
     def create_model():
         model = Sequential()
         model.add(Embedding(input_dim=vocab_size, output_dim=100,__
      →input_length=maxlen))
         model.add(Conv1D(128, 5, activation='relu'))
         model.add(MaxPooling1D(pool_size=2))
         model.add(Flatten())
         model.add(Dropout(0.2))
         model.add(Dense(5, activation='softmax'))
         model.compile(optimizer=Adam(learning_rate=0.001),__
      ⇔loss='categorical_crossentropy', metrics=['accuracy'])
         return model
     # 4. K-Fold Cross-Validation
     n_folds = 5
     kfold = KFold(n_splits=n_folds, shuffle=True)
     fold_no = 1
     acc_per_fold = []
     # Lists to store average accuracies across folds
     avg_train_acc = []
```

```
avg_val_acc = []
for train, test in kfold.split(X_resampled, y_resampled):
    model = create_model()
    history = model.fit(X_resampled[train], y_resampled[train],__
 ⇔validation_split=0.1, epochs=10, batch_size=32, verbose=0)
    avg train acc.append(history.history['accuracy'])
    avg_val_acc.append(history.history['val_accuracy'])
    scores = model.evaluate(X_resampled[test], y_resampled[test], verbose=0)
    acc_per_fold.append(scores[1] * 100)
    fold_no += 1
# 5. Performance Metrics
print(f'> Average Accuracy across {n_folds}-folds: {np.mean(acc_per_fold):.2f}%__
 # Average Training and Validation Accuracy for 5 folds
plt.plot(np.mean(avg_train_acc, axis=0), label='Average Training Accuracy')
plt.plot(np.mean(avg_val_acc, axis=0), label='Average Validation Accuracy')
plt.title('Average Training and Validation Accuracy across 5 folds')
plt.xlabel('Epoch')
plt.ylabel('Accuracy')
plt.legend()
plt.show()
# Confusion Matrix
y_pred = model.predict(X_resampled)
y_pred_classes = np.argmax(y_pred, axis=1)
y_true_classes = np.argmax(y_resampled, axis=1)
cm = confusion_matrix(y_true_classes, y_pred_classes)
sns.heatmap(cm, annot=True, fmt='g')
plt.show()
# Classification Report
print(classification_report(y_true_classes, y_pred_classes,__
  →target_names=list(y_dict.keys())))
/Users/nekdilkhan/miniforge3/lib/python3.12/site-
packages/keras/src/layers/core/embedding.py:90: UserWarning: Argument
`input_length` is deprecated. Just remove it.
  warnings.warn(
> Average Accuracy across 5-folds: 97.19% (+/- 0.41%)
```



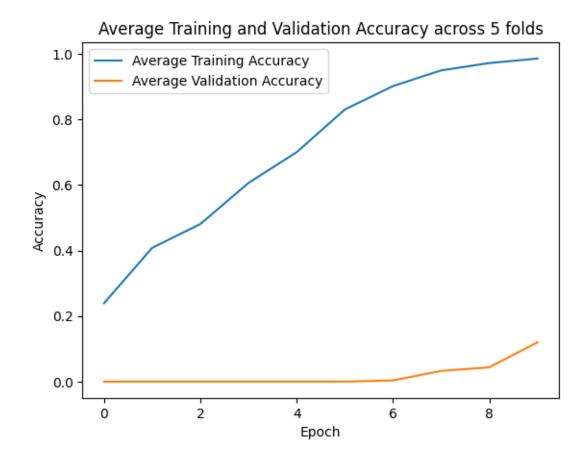
490/490 2s 4ms/step



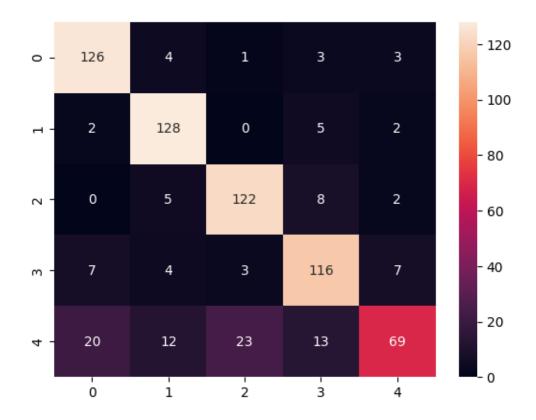
		precision	recall
f1-score	support		
	Navigation and Interaction Problems (NAV)	1.00	1.00
1.00	3136		
	Input and Control Issues (INPUT)	1.00	1.00
1.00	3136		
Compa	atibility with Assistive Technologies (CAT)	0.99	1.00
0.99	3136		
	UI Accessibility Issues (UI)	1.00	0.99
0.99	3136		
Audio and	d Visual Accessibility issues (AUDIOVISUAL)	0.99	0.98
0.98	3136		
	accuracy		
0.99	15680		
	macro avg	0.99	0.99
0.99	15680		
	weighted avg	0.99	0.99
0.99	15680		

```
[4]: #CNN manual annoated dataset dataset, undersampling
     import pandas as pd
     import numpy as np
     import re
     from nltk.stem import WordNetLemmatizer
     from sklearn.model_selection import KFold
     from tensorflow.keras.preprocessing.text import Tokenizer # Updated import
     from tensorflow.keras.preprocessing.sequence import pad_sequences # Updated_
      \hookrightarrow import
     from tensorflow.keras.models import Sequential # Updated import
     from tensorflow.keras.layers import Embedding, Conv1D, MaxPooling1D, Flatten,
      →Dropout, Dense # Updated import
     from tensorflow.keras.optimizers import Adam # Updated import
     from sklearn.metrics import confusion matrix, classification report
     import matplotlib.pyplot as plt
     import seaborn as sns
     from imblearn.under_sampling import RandomUnderSampler
     # 2. Undersample to balance classes
     undersampler = RandomUnderSampler(random_state=42)
     X_resampled, y_resampled = undersampler.fit_resample(X, y)
     # 3. Define the CNN model
     def create_model():
        model = Sequential()
         model.add(Embedding(input_dim=vocab_size, output_dim=100,__
      →input_length=maxlen))
         model.add(Conv1D(128, 5, activation='relu'))
         model.add(MaxPooling1D(pool_size=2))
         model.add(Flatten())
         model.add(Dropout(0.2))
         model.add(Dense(5, activation='softmax'))
         model.compile(optimizer=Adam(learning_rate=0.001),__
      →loss='categorical_crossentropy', metrics=['accuracy'])
         return model
     # 4. K-Fold Cross-Validation
     n folds = 5
     kfold = KFold(n_splits=n_folds, shuffle=True)
     fold no = 1
     acc_per_fold = []
     # Lists to store average accuracies across folds
     avg_train_acc = []
     avg_val_acc = []
```

```
for train, test in kfold.split(X_resampled, y_resampled):
    model = create_model()
    history = model.fit(X_resampled[train], y_resampled[train],__
  ⇒validation_split=0.1, epochs=10, batch_size=32, verbose=0)
    avg train acc.append(history.history['accuracy'])
    avg_val_acc.append(history.history['val_accuracy'])
    scores = model.evaluate(X_resampled[test], y_resampled[test], verbose=0)
    acc_per_fold.append(scores[1] * 100)
    fold_no += 1
# 5. Performance Metrics
print(f'> Average Accuracy across {n_folds}-folds: {np.mean(acc_per_fold):.2f}%__
 \rightarrow (+/- {np.std(acc_per_fold):.2f}%)')
# Average Training and Validation Accuracy for 5 folds
plt.plot(np.mean(avg_train_acc, axis=0), label='Average Training Accuracy')
plt.plot(np.mean(avg_val_acc, axis=0), label='Average Validation Accuracy')
plt.title('Average Training and Validation Accuracy across 5 folds')
plt.xlabel('Epoch')
plt.ylabel('Accuracy')
plt.legend()
plt.show()
# Confusion Matrix
y_pred = model.predict(X_resampled)
y_pred_classes = np.argmax(y_pred, axis=1)
y_true_classes = np.argmax(y_resampled, axis=1)
cm = confusion_matrix(y_true_classes, y_pred_classes)
sns.heatmap(cm, annot=True, fmt='g')
plt.show()
# Classification Report
print(classification_report(y_true_classes, y_pred_classes,__
  →target_names=list(y_dict.keys())))
/Users/nekdilkhan/miniforge3/lib/python3.12/site-
packages/keras/src/layers/core/embedding.py:90: UserWarning: Argument
`input_length` is deprecated. Just remove it.
 warnings.warn(
> Average Accuracy across 5-folds: 47.59% (+/- 2.37%)
```



22/22 0s 11ms/step



		precision	recall
f1-score	support		
	Navigation and Interaction Problems (NAV)	0.81	0.92
0.86	137		
	Input and Control Issues (INPUT)	0.84	0.93
0.88	137		
Compa	tibility with Assistive Technologies (CAT)	0.82	0.89
0.85	137		
	UI Accessibility Issues (UI)	0.80	0.85
0.82	137		
Audio and	Visual Accessibility issues (AUDIOVISUAL)	0.83	0.50
0.63	137		
	accuracy		
0.82	685		
	macro avg	0.82	0.82
0.81	685		
	weighted avg	0.82	0.82
0.81	685		

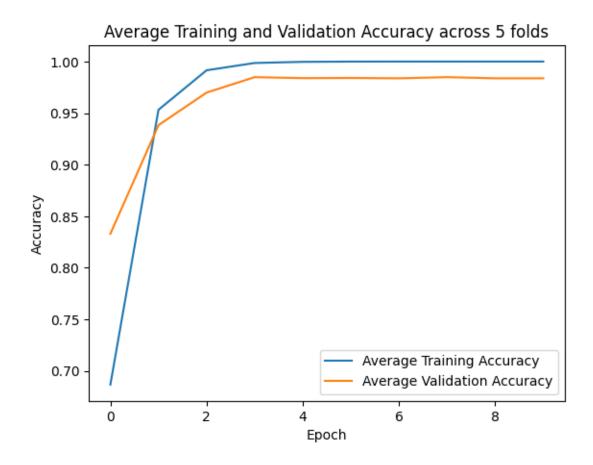
```
[5]: import pandas as pd
     import numpy as np
     import re
     from nltk.stem import WordNetLemmatizer
     from sklearn.model_selection import KFold
     from tensorflow.keras.preprocessing.text import Tokenizer
     from tensorflow.keras.preprocessing.sequence import pad_sequences
     from tensorflow.keras.models import Sequential
     from tensorflow.keras.layers import Embedding, Conv1D, MaxPooling1D, Flatten,
      →Dropout, Dense
     from tensorflow.keras.optimizers import Adam
     from sklearn.metrics import confusion_matrix, classification_report
     import matplotlib.pyplot as plt
     import seaborn as sns
     from imblearn.over_sampling import RandomOverSampler
     from sklearn.preprocessing import label_binarize # Correct import
     from sklearn.metrics import roc curve, auc
     # 2. Oversample to balance classes
     oversampler = RandomOverSampler(random_state=42)
     X_resampled, y_resampled = oversampler.fit_resample(X, y)
     # 3. Define the CNN model
     def create_model():
         model = Sequential()
         model.add(Embedding(input_dim=vocab_size, output_dim=100,__
      →input_length=maxlen))
         model.add(Conv1D(128, 5, activation='relu'))
         model.add(MaxPooling1D(pool size=2))
         model.add(Flatten())
         model.add(Dropout(0.2))
         model.add(Dense(5, activation='softmax'))
         model.compile(optimizer=Adam(learning rate=0.001),
      ⇔loss='categorical_crossentropy', metrics=['accuracy'])
         return model
     # 4. K-Fold Cross-Validation
     n folds = 5
     kfold = KFold(n splits=n folds, shuffle=True)
     fold no = 1
     acc_per_fold = []
     # Lists to store average accuracies across folds
     avg_train_acc = []
     avg_val_acc = []
```

```
for train, test in kfold.split(X_resampled, y_resampled):
    model = create_model()
    history = model.fit(X_resampled[train], y_resampled[train],__
 ⇒validation_split=0.1, epochs=10, batch_size=32, verbose=0)
    avg train acc.append(history.history['accuracy'])
    avg_val_acc.append(history.history['val_accuracy'])
    scores = model.evaluate(X_resampled[test], y_resampled[test], verbose=0)
    acc_per_fold.append(scores[1] * 100)
    fold_no += 1
# 5. Performance Metrics
print(f'> Average Accuracy across {n_folds}-folds: {np.mean(acc_per_fold):.2f}%__
 \rightarrow (+/- {np.std(acc_per_fold):.2f}%)')
# Average Training and Validation Accuracy for 5 folds
plt.plot(np.mean(avg_train_acc, axis=0), label='Average Training Accuracy')
plt.plot(np.mean(avg_val_acc, axis=0), label='Average Validation Accuracy')
plt.title('Average Training and Validation Accuracy across 5 folds')
plt.xlabel('Epoch')
plt.ylabel('Accuracy')
plt.legend()
plt.show()
# 6. ROC curve
y pred = model.predict(X resampled)
y_true_classes = np.argmax(y_resampled, axis=1)
y_pred_probs = model.predict(X_resampled)
# Binarize the true labels for ROC
y_true_binarized = label_binarize(y_true_classes, classes=[0, 1, 2, 3, 4, 5, 6, __
 →7])
# Initialize variables to store fpr, tpr, and auc for each class
fpr = dict()
tpr = dict()
roc_auc = dict()
# Calculate ROC curve and AUC for each class
for i in range(5):
    fpr[i], tpr[i], _ = roc_curve(y_true_binarized[:, i], y_pred_probs[:, i])
    roc_auc[i] = auc(fpr[i], tpr[i])
# Plot ROC curves for each class
plt.figure()
colors = ['red', 'blue', 'green', 'orange', 'purple']
```

```
for i, color in zip(range(5), colors):
   plt.plot(fpr[i], tpr[i], color=color, lw=2, label=f'ROC curve (class {i})_u
 ⇔(area = {roc_auc[i]:.2f})')
plt.plot([0, 1], [0, 1], color='navy', lw=2, linestyle='--')
plt.xlim([0.0, 1.0])
plt.ylim([0.0, 1.05])
plt.xlabel('False Positive Rate')
plt.ylabel('True Positive Rate')
plt.title('Receiver Operating Characteristic for Multiclass')
plt.legend(loc='lower right')
plt.show()
# 7. Confusion Matrix
y_pred_classes = np.argmax(y_pred, axis=1)
cm = confusion_matrix(y_true_classes, y_pred_classes)
sns.heatmap(cm, annot=True, fmt='g')
plt.show()
# 8. Classification Report
print(classification_report(y_true_classes, y_pred_classes,__
 ⇔target_names=list(y_dict.keys())))
```

```
/Users/nekdilkhan/miniforge3/lib/python3.12/site-
packages/keras/src/layers/core/embedding.py:90: UserWarning: Argument
`input_length` is deprecated. Just remove it.
warnings.warn(
```

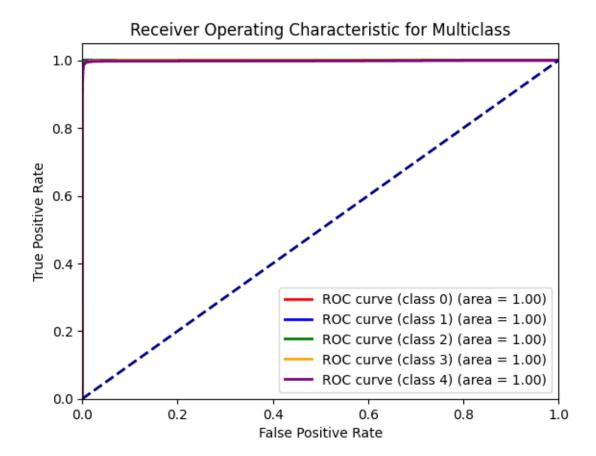
> Average Accuracy across 5-folds: 97.16% (+/- 0.12%)



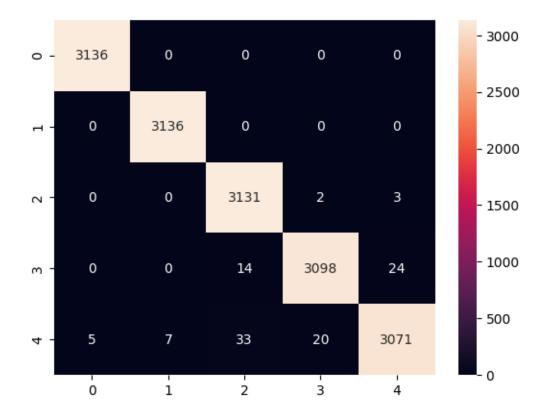
 490/490
 2s 4ms/step

 490/490
 2s 4ms/step

[6]:



[7]:



		precision	recall	
f1-score	support			
	Navigation and Interaction Problems (NAV)	1.00	1.00	
1.00	3136			
	Input and Control Issues (INPUT)	1.00	1.00	
1.00	3136			
Compa	atibility with Assistive Technologies (CAT)	0.99	1.00	
0.99	3136			
	UI Accessibility Issues (UI)	0.99	0.99	
0.99	3136			
Audio and	d Visual Accessibility issues (AUDIOVISUAL)	0.99	0.98	
0.99	3136			
	accuracy			
0.99	15680			
	macro avg	0.99	0.99	
0.99	15680			
	weighted avg	0.99	0.99	
0.99	15680			

```
[9]: import pandas as pd
     import numpy as np
     import re
     from nltk.stem import WordNetLemmatizer
     from sklearn.model_selection import KFold
     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, MaxPooling1D, Flatten,
      →Dropout, Dense
     from tensorflow.keras.optimizers import Adam
     from sklearn.metrics import confusion_matrix, classification_report
     import matplotlib.pyplot as plt
     import seaborn as sns
     from imblearn.over_sampling import RandomOverSampler
     from sklearn.preprocessing import label_binarize
     from sklearn.metrics import roc_curve, auc
     # Assuming X and y are defined already, and data preprocessing is done
     # Example: X = \ldots, y = \ldots
     # 2. Oversample to balance classes
     oversampler = RandomOverSampler(random_state=42)
     X_resampled, y_resampled = oversampler.fit_resample(X, y)
     # 3. Define the CNN model
     def create_model():
         model = Sequential()
         model.add(Embedding(input_dim=vocab_size, output_dim=100,__
      →input_length=maxlen))
         model.add(Conv1D(128, 5, activation='relu'))
         model.add(MaxPooling1D(pool_size=2))
         model.add(Flatten())
         model.add(Dropout(0.2))
         model.add(Dense(5, activation='softmax'))
         model.compile(optimizer=Adam(learning_rate=0.001),__
      ⇔loss='categorical_crossentropy', metrics=['accuracy'])
         return model
     # 4. K-Fold Cross-Validation
     n_folds = 5
     kfold = KFold(n_splits=n_folds, shuffle=True)
     fold_no = 1
     acc_per_fold = []
```

```
# Lists to store average accuracies and losses across folds
avg_train_acc = []
avg_val_acc = []
avg_train_loss = []
avg_val_loss = []
for train, test in kfold.split(X_resampled, y_resampled):
    model = create model()
    history = model.fit(X_resampled[train], y_resampled[train],__
 ⇒validation split=0.1, epochs=10, batch size=32, verbose=0)
    avg_train_acc.append(history.history['accuracy'])
    avg_val_acc.append(history.history['val_accuracy'])
    avg_train_loss.append(history.history['loss'])
    avg_val_loss.append(history.history['val_loss'])
    scores = model.evaluate(X_resampled[test], y_resampled[test], verbose=0)
    acc per fold.append(scores[1] * 100)
    fold no += 1
# 5. Performance Metrics
print(f'> Average Accuracy across {n_folds}-folds: {np.mean(acc_per_fold):.2f}%__
 \rightarrow (+/- {np.std(acc_per_fold):.2f}%)')
# 6. Plotting Training and Validation Accuracy and Loss in the same graph
def plot history():
    plt.figure(figsize=(12, 6))
    # Plot for model accuracy
    plt.subplot(1, 2, 1)
    plt.plot(np.mean(avg_train_acc, axis=0), label='Average Training Accuracy')
    plt.plot(np.mean(avg_val_acc, axis=0), label='Average Validation Accuracy')
    plt.title('Training and Validation Accuracy')
    plt.xlabel('Epoch')
    plt.ylabel('Accuracy')
    plt.legend()
    # Plot for model loss
    plt.subplot(1, 2, 2)
    plt.plot(np.mean(avg_train_loss, axis=0), label='Average Training Loss')
    plt.plot(np.mean(avg_val_loss, axis=0), label='Average Validation Loss')
    plt.title('Training and Validation Loss')
    plt.xlabel('Epoch')
    plt.ylabel('Loss')
    plt.legend()
```

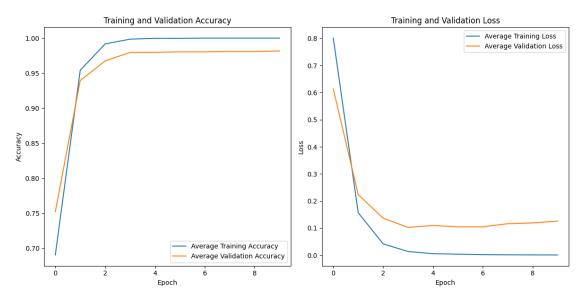
```
plt.tight_layout()
    plt.show()
# Plot the results
plot_history()
# 7. ROC curve
y_pred = model.predict(X_resampled)
y_true_classes = np.argmax(y_resampled, axis=1)
y_pred_probs = model.predict(X_resampled)
# Binarize the true labels for ROC
y_true_binarized = label_binarize(y_true_classes, classes=[0, 1, 2, 3, 4, 5, 6, __
 →7])
# Initialize variables to store fpr, tpr, and auc for each class
fpr = dict()
tpr = dict()
roc_auc = dict()
# Calculate ROC curve and AUC for each class
for i in range(5):
    fpr[i], tpr[i], _ = roc_curve(y_true_binarized[:, i], y_pred_probs[:, i])
    roc_auc[i] = auc(fpr[i], tpr[i])
# Plot ROC curves for each class
plt.figure()
colors = ['red', 'blue', 'green', 'orange', 'purple']
for i, color in zip(range(5), colors):
    plt.plot(fpr[i], tpr[i], color=color, lw=2, label=f'ROC curve (class {i})_u
 \Rightarrow(area = {roc_auc[i]:.2f})')
plt.plot([0, 1], [0, 1], color='navy', lw=2, linestyle='--')
plt.xlim([0.0, 1.0])
plt.ylim([0.0, 1.05])
plt.xlabel('False Positive Rate')
plt.ylabel('True Positive Rate')
plt.title('Receiver Operating Characteristic for Multiclass')
plt.legend(loc='lower right')
plt.show()
# 8. Confusion Matrix
y_pred_classes = np.argmax(y_pred, axis=1)
cm = confusion_matrix(y_true_classes, y_pred_classes)
sns.heatmap(cm, annot=True, fmt='g')
plt.title('Confusion Matrix')
plt.show()
```

```
# 9. Classification Report

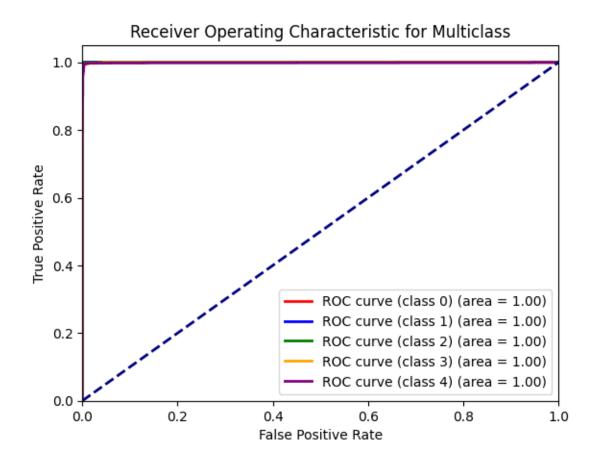
print(classification_report(y_true_classes, y_pred_classes, u_target_names=[f"Class {i}" for i in range(5)]))
```

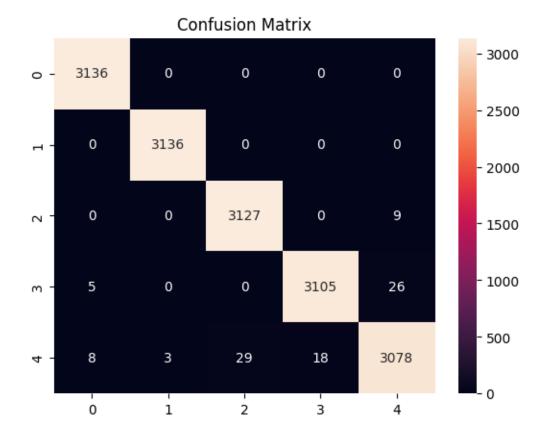
/Users/nekdilkhan/miniforge3/lib/python3.12/sitepackages/keras/src/layers/core/embedding.py:90: UserWarning: Argument `input_length` is deprecated. Just remove it. warnings.warn(

> Average Accuracy across 5-folds: 96.94% (+/- 0.41%)



490/490 2s 4ms/step 490/490 2s 4ms/step





	precision	recall	f1-score	support
Class 0	1.00	1.00	1.00	3136
Class 1	1.00	1.00	1.00	3136
Class 2	0.99	1.00	0.99	3136
Class 3	0.99	0.99	0.99	3136
Class 4	0.99	0.98	0.99	3136
accuracy			0.99	15680
macro avg	0.99	0.99	0.99	15680
weighted avg	0.99	0.99	0.99	15680