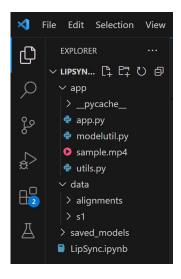
PROJECT MANUAL

Team Members

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Project Structure:



1)app folder contains the app.py which is used for building web site for our model it also contains the utils.py and modelutil.py. modelutil.py is used for loading the model and utils.py contains the functions used in website building.

- 2)Saved_models folder contains the models saved after the model training.
- 3)The LipSync.ipynb notebook will have the model which we are building using the Deep Learning.
- 4)data folder contains s1 and alignments folders. S1 folder contains the videos for the training and alignments contains the annotations of the video.

MODEL (LipSync.ipynb):

```
### Dispincion N

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| Importing Necessary Libraries

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```

In First cell we are importing all the necessary libraries used for the model building.

In second cell we are using a load_video(path) used for load video of the particular path , in this we are using the open cv library to capture the frames of the video and we are trimming the frames to capture the particularly lips region of the each frame and then we are normalising the frames for optimisation.

```
# Upsyncipynb X # Functions for text and video for Processing > Φ def load_alignmentspathaty>- Listpty:

# Upsyncipynb X # Functions for text and video for Processing > Φ def load_alignmentspathaty>- Listpty:

# Vocab = [x for x in "abcdefghijklamopqrstuvwyz'?1123456789"]

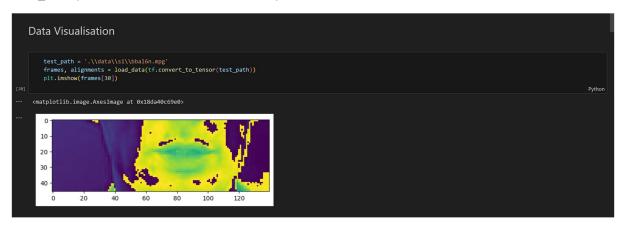
| Vocab = [x for x in "abcdefghijklamopqrstuvwyz'?1123456789"]

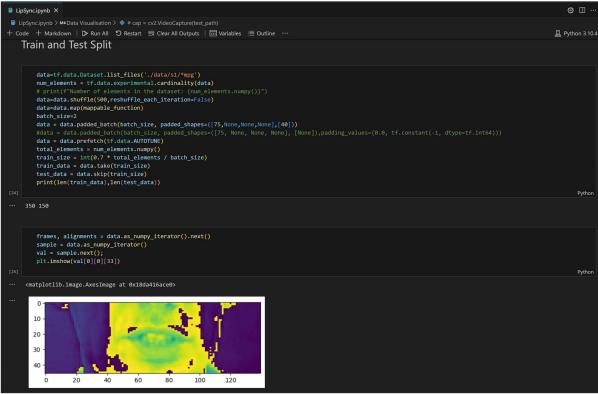
| Char_to_num = tf.keras_layers_stringLookup(vocabularyvvocab, oov_token="")
| num_to_char = tf.keras_layers_stringLookup(
| vocabularyvchar_to_num_ept_vocabulary(), oov_token="")
| num_to_char = tf.keras_layers_stringLookup(
| vocabularyvchar_to_num_ept_vocabulary(), oov_token="")
| print(
| f'The vocabulary is: (char_to_num_ept_vocabulary()) "
| f'(size = (char_to_num_vocabulary); size()))"
| print(
| f'The vocabulary is: ['', 'a', 'b', 'c', 'd', 'e', 'f', 'g', 'h', 't', 'j', 'k', 'l', 'm', 'n', 'o', 'p', 'q', 'r', 's', 't', 'u', 'w', 'w', 'x', 'y', 'z', ''', 'z', ''', 'z', ''', 'z', ''', 'z', ''', 'z', ''', 'z', 'z', ''', 'z', 'z',
```

We are using char to num and num to char functions ,as the model will take the inputs of numbers these are used in the further parts.

load_alignments(path) function is used for the loading of captions or annotations of the video with given path

load_data(path) is used to load video and captions.





We are splitting the data into two batches and training and testing with 70% and 30% of the data in batch.

```
model=Sequential()
model.add(Convolution3D(128,3,input_shape=(75,46,140,1),padding="same",activation="relu"))
model.add(Govolution3D(128,3,input_shape=(75,46,140,1),padding="same",activation="relu"))
model.add(Govolution3D(265,3,padding="same",activation="relu"))
model.add(Govolution3D(265,3,padding="same",activation="relu"))
model.add(MaxPooling3D(pool_size=(1,2,2)))
model.add(MaxPooling3D(pool_size=(1,2,2)))
model.add(Ginesistributed(Flatten()))
model.add(Ginesistributed(Flatten()))
model.add(Ginectional(LSTM(128,kernel_initializer="Orthogonal",return_sequences=True)))
model.add(Dropout(.5))
model.add(Oropout(.5))
model.add(Convolution3D(Same),return_sequences=True)))
model.add(Dense(128,activation="relu"))
model.add(Dense(128,activation="relu"))
model.add(Dense(cflar_to_num.vocabulary_size()+1, kernel_initializer="he_normal", activation="softmax"))
model.output_shape

Python
```

We are using CNN and RNN algorithms for building this model.

CTC loss is the loss function is used In compiling model it is more appropriate loss function for this model.

```
model.compile(optimizer=Adam(learning_rate=0.0001), loss=CTCLoss)
checkpoint_callback = ModelCheckpoint(os.path.join('saved_models','checkpoint'), monitor='loss', save_weights_only=True)
schedule_callback = Producetxample(test_data)
model.fit(train_data, validation_data=test_data, epochs=100, callbacks=[checkpoint_callback, schedule_callback, example_callback])

Python
```

After 100 epochs the model will be saved in the saved_models folder with name checkpoint callbacks is used for optimising the model just as kind of early stopping.

```
Testing

model.load_weights('saved_models/checkpoint')

test= test_data.as_numpy_iterator()
sample = test.next()
y_pred = model.predict(sample[0])

print('~'*100, 'Original Text')
[ff.strings.reduce_join([num_to_char(word) for word in sentence]) for sentence in sample[1]]
decoded = tf.keras_backend.ctc_decode(y_pred, input_length=[75,75], greedy=True)[0][0].numpy()
print('~'*100, 'Predicted Text')
[ff.strings.reduce_join([num_to_char(word) for word in sentence]) for sentence in decoded]

[1]
```

Original:

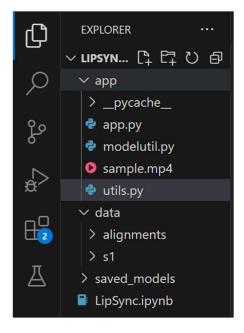
Predicted:

WEB PAGE BUILDING USING STREAMLIT:

We are setting the layout of the webpage as wide using the streamlit library then we are setting up the sidebar with a image and some texts.

Next we are having two coulms in which one column is for selecting the video and diaplaying the selected video. Also as our video data is in form of mpg we are converting the mpg to mp4 with use of the ffmpeg library as stramlit library doesn't support playing the mpg format video.

Another column is for displaying the predicted text from video.



The video will be saved with sample.mp4 an this what we are playing in website by retrieving.

```
LipSync.ipynb
                  🕏 арр.ру
                                 modelutil.py ×
app > ♦ modelutil.py > ♦ load_model
      from keras.layers import Conv3D, LSTM, Dense, Dropout, Bidirectional, MaxPool3D, Activation, TimeDistributed, Flatten
      def load_model() -> Sequential:
          model = Sequential()
          model.add(Conv3D(128, 3, input_shape=(75,46,140,1), padding='same'))
          model.add(Activation('relu')'
          model.add(MaxPool3D((1,2,2)))
          model.add(Conv3D(256, 3, padding='same'))
          model.add(Activation('relu'))
          model.add(MaxPool3D((1,2,2)))
          model.add(Conv3D(75, 3, padding='same'))
          model.add(Activation('relu'))
          model.add(MaxPool3D((1,2,2)))
          model.add(TimeDistributed(Flatten()))
          model.add(Bidirectional(LSTM(128, kernel_initializer='Orthogonal', return_sequences=True)))
          model.add(Dropout(.5))
          model.add(Bidirectional(LSTM(128, kernel_initializer='Orthogonal', return_sequences=True)))
          model.add(Dropout(.5))
          model.add(Dense(41, kernel_initializer='he_normal', activation='softmax'))
          model.load_weights(os.path.join('...', 'saved_models', 'checkpoint'))
          return model
```

These are the same functions we are used in model building.

WEBSITE:

```
Warning: PowerShell detected that you might be using a screen reader and has disabled PSReadLine for compatibility purposes. If you want to re-enable it, run 'Import-Module PSReadLine'.

PS C:\Users\indup\OneDrive - vitap.ac.in\Desktop\Project_Development_Phase\LipSyncStudio> cd app
PS C:\Users\indup\OneDrive - vitap.ac.in\Desktop\Project_Development_Phase\LipSyncStudio\app> streamlit run app.py

You can now view your Streamlit app in your browser.

Local URL: http://localhost:8501
Network URL: http://localhost:8501
Network URL: http://localhost:8501 is tensorflow/core/platform/cpu_feature_guard.cc:182] This Tensorflow binary is optimized to use available CPU instructions in performance-critical operations.

To enable the following instructions: SSE SSE2 SSE3 SSE4.1 SSE4.2 AVX AVX2 AVX512_VNNI FMA, in other operations, rebuild Tensorflow with the appropria
```

This is how we will run the website

